Modelling the effects of lymph node swelling on T-cell response

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Abstract

Swelling of the lymph nodes is commonly observed during the adaptive immune response, yet its impacts on T cell trafficking and subsequent immune response are not well known. To better understand the effect of macro-scale alterations in the lymph node, we developed an agent-based model of the lymph node paracortex, describing T cell trafficking and response to antigen-presenting dendritic cells alongside swelling-induced changes in T cell recruitment and egress, and regulation of expression of egress-modulating T cell receptor Sphingosine-1-phosphate receptor-1. Validation of the model was achieved with in-silico replication of a range of published in-vivo and cell culture experiments. Analysis of CD4+ and CD8+ effector T cell response under varying swelling conditions showed that paracortical swelling aided initial T cell activation but could inhibit subsequent effector CD8+ T cell production if swelling occurs too early in the T cell proliferative phase. This was primarily due to changes in effector T cell egress rather than changes in T cell and Dendritic cell interactions. Temporarily extending retention of newly differentiated effector T cells, mediated by Sphingosine-1-phosphate receptor-1 expression, mitigated the effects of early paracortical swelling. These results suggest that targeting the timing of lymph node swelling and temporary effector T cell retention may offer new ways to modulate effector TC responses in for example, immuno-suppressed patients or to optimise vaccine design.

Author summary

Within lymph nodes, interaction of T cells and antigen-presenting cells play a crucial role in initiating the adaptive immune response, resulting in effector T cells that travel to the infection site. Accompanying lymph node swelling is commonly observed, yet the impact on T cell trafficking and subsequent immune response is not well understood. We developed an agent-based lymph node model, describing immune response-induced expansion and changes
in T cell recruitment and egress. We included regulation of T cell expression of Sphingosine-1-phosphate receptor-1 due to the important role in T cell trafficking. We found that although swelling aids T cell activation, too early an increase in paracortical volume can hinder CD8+ effector T cell response. This was primarily due altered in effector T cell egress rather than T cell interactions. Temporary extension of Sphingosine-1-phosphate receptor-1 downregulation on newly differentiated effector T cells greatly increased overall effector T cell output and could counteract loss in effector T cell production due to early swelling. Increased T cell retention during initial lymph node shutdown did not produce the same effect. Our findings suggest targeting the timing of lymph node swelling and temporary effector T cell retention may offer new ways to manipulate immune response.

**Introduction**

The lymphatic system is a network of organs and lymphatic vessels (LVs) that maintains fluid balance in the body and also delivers crucial antigen information to lymph nodes (LNs) for initiation of adaptive immunity. A successful immune response relies not only on the interaction of different immune cell types, but also on the maintenance of an appropriate physical environment in the LNs to facilitate those interactions. LNs contain specific compartments populated by T cells (TCs), B cells, fibroreticular cells (FRCs), and lymphatic endothelial cells (LECs) [1,2]. When antigens are presented (either suspended in lymph or as captured by incoming antigen presenting cells), the LNs’ physical environment changes over time. Swelling of LNs is a well-known consequence of antigen presentation, but the effects of swelling on the processes crucial for adaptive immunity are not well understood.

The mechanism driving LN swelling in response to antigens can be attributed to more than one factor, for example, Dendritic Cell (DCs) presence, B cell signalling and trapping of non-
activated TCs in the LNs [3-6]. During oedema, where peripheral lymphatic fluid accumulates, LN swelling is not routinely observed; therefore, swelling due to increased afferent fluid loading alone is unlikely. Regardless of the trigger, within the first 2 days, TC exit rate falls (sometimes called “LN shutdown”), blood flow to the LN increases, and inflammatory signalling results in a 3-5 fold increase in TC recruitment via high endothelial venules (HEV) [7-10]. LN mass increases 2-5-fold, accompanied by a similar increase in cellularity; in the first 48-96 hours, and the FRCs elongate to accommodate the increase in LN size [6,11,12]. Proliferation of stromal cells lags behind the increase in immune cells, but subsequent proliferation of LECs and FRCs allow maintenance of the LNs architecture during further expansion [5,12,13]. The blood vessels of the LNs also grow, increasing blood vessel volume roughly proportional to the overall volume of the LNs, accompanied by further TC recruitment [4,9,14].

Between 2- and 5-days post-immunisation, the number of antigen-presenting DCs in LNs peaks, TC activation and proliferation is underway and TC egress increases 3-6 fold [5,6,15,16]. Expansion of the medullary and subcapsular sinus (SCS) areas has also been observed, which aids in the increase in TC egress rate [17]. Recruitment of TCs then declines, HEV, FRC and TC proliferation subsides, remaining effector TCs may undergo apoptosis and the LNs return to their baseline volume [14]. Following a primary response, FRC, LEC and blood vessel endothelial cell numbers remain elevated within denser LNs structures for at least one month [13].

To pass through these phases, the underlying antigenic stimulus must be sufficient to trigger lymphocyte retention, activation and proliferation, while a successful immune response occurs when TCs further differentiate into effector cells and migrate out of the LNs [18].
Migrating DCs cross the SCS floor in a chemokine and integrin-aided fashion and migrate into the paracortex [19]. TCs and B Cells mainly enter the LNs by transmigrating from blood vessels in the paracortex [20]. Typically, 1 in 10,000 naïve TCs express a complementary TC receptor to the antigen fragment presented by DCs within a MHCI (to CD8+ TCs) or MHCII (CD4+ TCs) molecule [21,22]. Naive cognate TCs initially make short contacts with DCs, but after around 8 hours, progress to longer interactions (>1hr) before returning to short interactions as TCs activate and divide [23]. With sufficient affinity and stimuli, TCs undergo activation, and secrete inflammatory and activation-facilitating cytokines [24]. Some TCs differentiate into effector TCs and up-regulate the Sphingosine-1-phosphate receptor-1 (S1P1r), facilitating egress [25]. An increasing proportion of TCs differentiate into memory cells [26,27]. Proliferation of CD8+ TCs can continue independent of further stimuli, possibly with an impaired memory cell response [28,29]. Contrastingly, CD4+ TCs are more dependent on sustained inflammatory stimulation for continued differentiation [30]. As infectious signals subside, remaining effector TCs undergo apoptosis while memory cells go into circulation [31].

Throughout these processes, the egress of TCs from the LNs is modulated by Sphingosine-1-Phosphate (S1P) and chemokine signalling axes. After entering the LNs, TCs initially express S1P1r at very low levels but begin re-expressing S1P1r after 2 hours [32,33]. TCs exit the LNs by probing and subsequently entering cortical sinuses in the paracortex or the interface with the medulla, aided by chemotaxis as both destinations contain higher S1P concentrations (Fig 1A) [34,35]. During inflammation, TC S1P1r expression is reciprocally regulated by CD69, an early TC activation marker that can be upregulated in TCs by the presence of inflammatory mediators, contributing to the initial variable decrease of up to 80% in TC egress, termed LN shutdown, and later in the retention of activated TCs [10,36].
**Figure 1: Model geometry and structure.** (A) LN structure displaying the pathway of arriving lymphatic fluid containing agDCs. (B) The model contains a spherical paracortex. TCs enter in the centre and exit near the interface with the medulla and SCS. The paracortex radius expands as a function of TCs present. (C) At the microscale, TCs move to adjacent grid-compartments (6 of 8 possible movements shown), interact with neighbouring agents and are influenced by the grid-compartment properties, which are updated each time-step. Microscale signals can affect TC recruitment, TC exit and stimulation via interaction with DCs. (D) Interactions between micro and macro scales.

The exploration of potential roles for LN swelling in adaptive immunity is limited by the range of viable experiments that can target specific mechanisms or components while also tracking relevant outcomes in real time. Mathematical models allow time-specific manipulation and tracking of experimentally unobtainable variables, with outcomes helping to fill knowledge gaps and optimise experimental design. We chose to develop a model that could describe both macroscale geometric changes and the microscale TC and DC interactions that drive LN swelling. Agent Based Models (ABMs) can model cells as discrete agents, allowing differing internal states that capture biological variance. By modelling the interaction of thousands of agents following probabilistic rules and behaving individually, it is possible to capture emergent behaviours. ABMs also allow for spatially realistic descriptions of HEV location, afferent lymphatics and TC exit areas into the medullary sinuses. These points are critical for an accurate description of cell transit, as well as for future integration with an antigen and chemokine mass transport model. The FRC network was not incorporated, as previous lattice-based models have shown minimal impact on TC
cell contacts and transit \[37-39\], and random-walk behaviour of TCs is observed *in-vivo* regardless of stromal cell contribution \[40-43\].

Several ABMs of fixed-volume LNs have been developed to analyse the influence of the micro-scale signal interaction level. These models have provided insight relevant to vaccine design, for example to investigate the effect of antigenic peptide separation from MHC molecules on TC activation, influential aspects of TC-DC interaction, and effector TC or memory cell production \[44-48\]. Similar models that include paracortical expansion assume a constant TC occupancy of the total paracortical volume, which may neglect the effects of crowding on TC migration \[49-51\]. As our focus is on macroscale changes and the resulting physical impact, rather than refining descriptions of intra-cellular signalling cascades, allowing changes in cellular crowding is particularly important due to the potential impact on egress rates and future implementations of pressure and flow. Simulations integrating a fixed-volume hybrid lattice-based model and a continuous model allowed incorporation of chemokine diffusion with TC, B cell and DC interactions in the LNs. This showed that both early antigen removal and regulation of TC exit affected the balanced system dynamics, indicating that macroscale swelling is likely to significantly affect micro-scale TC activity \[52\]. Indeed, models of lymph flow through the LNs suggest that altered oncotic and medulla pressure alone would dramatically alter cell, cytokine and chemokine distributions \[53,54\].

In summary, the adaptive immune response involves careful trafficking and coordination of immune cell movements in the LNs, suggesting that swelling of LNs is likely to significantly impact the adaptive response at the micro-scale. To investigate this hypothesis, a computational ABM was developed that incorporates TC and DC dynamics, swelling of LNs and changes in TC recruitment and egress. The model was first validated against available
experimental data. The results suggest an important role for swelling of LNs in TC population dynamics.

Materials and methods

ABM Geometry

The paracortex is modelled as a sphere with an initial radius of $R_0=200\mu$m, derived from confocal images [2,55]. To reduce computational expense, geometric symmetry is assumed so that only one-half of the total spherical geometry is modelled. The modelling domain is divided into cuboid grid compartments, with an edge length of $6\mu$m (Fig 1C). For each grid compartment, we track which area of the paracortex is represented, such as ‘exit’, ‘boundary’ or ‘outside’. Paracortical swelling or contraction is achieved by changing the type of region that each grid compartment represents, while maintaining entry and exit areas defined as a percentage of the outer radius (Fig 2A, Fig A in S1 File). To model swelling, we first collated and assessed information from murine experiments regarding changes in LN mass and volume, TC counts, structural cells counts, migrating DC counts, TC recruitment and TC egress following the application of an antigenic stimulus (Fig 2A) [4,6,7,14,15]. Based on this data, we calculate paracortical volume as a sigmoidal function of TCs present (Fig 2A).

Figure 2: Modelling the changes in paracortical volume, TC recruitment and TC expression of S1P$_1$r. A) Previously published experimental data from studies of multiple influences on LN swelling were used to inform the selection and parameterization of a sigmoidal function to drive LN swelling based on TC numbers. B) TC recruitment is driven by initial antigenic signal (eq. 2), with parameterization based on in-vivo data, and a proportional increase due to expansion of HEVs. C) S1P$_1$r expression mediated the retention
of (i) naïve TCs post-transmigration from the blood, (ii) naïve TCs during LN shutdown and
(iii) activated and early effector TCs.

**TC recruitment**

Under baseline conditions, TC recruitment rate is specified as 2000 TCs/hour, with naive TC
transit time ($T_{res}$) defined between 6-24 hours and the TC-to-compartment ratio assumed
constant (1.1 in S1 File). In accordance with HEV images, 90% of TCs enter at 'HEV entry'
compartments designated as the inner half of the paracortical radius [56]. The volume of
these entry grids also corresponds to the blood vessel volume ($V_B$), which varies
proportionally with the overall change in paracortical volume [4,53]. Remaining TCs enter
via the SCS interface at compartments adjacent to the afferent half of the external surface.
The TC recruitment rate ($T_{in}$), is proportional to $V_B$, starting with a predefined baseline value.
Acute recruitment changes due to inflammation-induced signalling cascades at the HEVs are
represented by an inflammatory index, $I_F$ (Eq 1). This index impacts TC influx when the
antigenic presence (sum of MHCII) rises above the trigger threshold, $T_1$, the minimal number
of DCs required to elicit a response [45]. The value of $I_F$ increases proportionally by a
recruitment factor ($R_F$) with antigenic presence up to a cap threshold, $T_2$, where the
maximum inflammation-induced TC recruitment is reached regardless of further stimuli (Eq.
2). TC influx is therefore defined as:

$$T_{in}(t) = \frac{N_T}{T_{res}} I_F(t) V_B(t)$$  \hspace{1cm} (1)$$

Where $N_T$ is the initial number of TCs present, $T_{res}$ is the naïve TC transit time, $I_F$ is the
inflammatory index, and $V_B$ is the normalised blood vessel volume. Threshold values for $T_1$,
T2 and RF were estimated from TC recruitment rate changes due to inflammation in the initial few days, whilst considering the changes due to HEV growth and agDC number present [6-10,15,57]. The inflammatory index $I_F$ is calculated as:

$$I_F(t) = \begin{cases} 
1 & \sum_{n=1}^{N_{DC}} MHC_{II}(t) \leq T1 \\
1 + RF \sum_{n=1}^{N_{DC}} MHC_{II}(t) & T1 \leq \sum_{n=1}^{N_{DC}} MHC_{II}(t) \leq T2 \\
1 + RF T2 & \sum_{n=1}^{N_{DC}} MHC_{II}(t) \leq T2 
\end{cases}$$

(2)

Where Recruitment Factor $RF$ is the increase in recruitment rate, $N_{DC}$ is the number of agDCs present, and $MHC_{II}$ is the sum of MHCII carried by each agDC.

**TC egress and S1P1r expression**

We alter relative TC expression of S1P1r ($SP$), from the default value of 1, under three conditions (Fig 2C). Firstly, following TC entry into the paracortex from the blood, S1P1r remains down-regulated ($SP_{in}=0.1$) for 45-180 minutes, before re-expressing due to low paracortical S1P concentration [33]. A mechanism of ‘LN shutdown’ is included by down-regulating S1P1r ($SP_{inflam}=0.4$) on all TCs when antigenic presence is detected. Summation of MHCII present is used as an antigenic signal, set initially to trigger after 6 hours of baseline simulation. Finally, activation-induced TC S1P1r down-regulation is included by applying the largest decrease in S1P1r expression when TCs first become activated ($SP_{act}=0.01$), then increasing S1P1r expression as TCs differentiate into effector TCs ($SP_{early}=0.4$), and further increasing expression when effector TCs undergo $\geq 8$ divisions ($SP_{late}=1$) [32, 58, 59].
**TC and DC motility and interaction**

DCs are modelled as 6μm radius spheres. To reflect their long dendrites, DCs can interact with TCs within a two-grid radius, up to a user-defined maximum number of TCs at one time, defined by parameter, $B_{\text{max}}$ (Fig A in S1 File). Interaction times are drawn from uniform probability distributions, with brief 3-minute interactions for non-cognate TCs ($T_{\text{NC}}$).

Cognate TCs initially undergo short ($T_{\text{short}}$) interactions of 10-15 minutes, then proceed to longer interactions ($T_{\text{long}}$) of 50-70 minutes. Each agDC presents a decaying MHCI and MHCII signal with half-lives $\text{MHCI}_{1/2}$ and $\text{MHCII}_{1/2}$, obtained from *in-vitro* labelling of MHC molecules presented on DCs [60-63].

TCs are modelled as spheres of volume 150 μm$^3$ with radius 3.3μm that initially occupy 60% of the total paracortex volume [64]. The starting TC population is composed of 70% helper TCs (CD4+) and 30% cytotoxic TCs (CD8+) [65]. At each timestep ($\Delta t=20s$), TCs are permitted to move one grid length to an available neighbouring grid compartment, moving with respect to probability, $\beta$, and where availability is governed by crowding parameter, $\gamma$.

The proportion of cognate TCs recruited each timestep ($F_{\text{cog}}$) was obtained from reported *in-vivo* frequency of antigen-specific TCs [22].

Cognate TCs gain 'stimulation' ($S$) during interactions with agDCs at rate $K_s$, proportional to MHC presented, while losing stimulation at rate $\lambda_S$ (Fig A in S1 File), similar to the methods of [46, 50, 66]. The probability of subsequent activation and differentiation is determined as a sigmoidal function of accumulated stimulation. The functions were parameterised such that CD8$^+$ TCs require more accumulated stimulation to activate with the same probability as
CD4+ TCs. However, if the DC is 'licenced', which occurs post-interaction with an activated CD4+ TC, CD8+ TC and CD4+ TC stimulation requirements are equal. This is to reflect facilitated CD8+ activation as a result of activated CD4+ induced production of cytokines [67]. Proliferation is possible every 11 and 9 (+/-1) hours for CD4+ TCs and CD8+ TCs, respectively, while differentiation into effector or memory TCs is only possible after undergoing 4 proliferations [68-71]. A higher level of stimulation is required for CD4+ TCs to differentiate with the same probability as CD8+ TCs. The fraction of effector TCs that differentiate into memory TCs increases from 0.01 to 0.04 as TCs progress from ‘early effectors’ (< 8 proliferations) to ‘late effectors’ [72]. See S1 File for full rules and parameter reference Tables A&B.

Computation

The ABM was built in RepastSimphony (http://repast.sourceforge.net) as a class-based model (Fig B in S1 File) written in Java with repeated rules each timestep (Fig 3). Further UML-based descriptions are available within S2 File. The model can be visualised within a GUI (Fig 4). The Imperial College High Performance Computing cluster was used to carry out batch simulations. Data analysis was carried out in Matlab. Complete code is available at github.com/johnsara04/paracortex_model_johnson19.

Figure 3: Structure of the code to set up and run the model. A) The model is initiated in the absence of stimulus, then capacity for paracortical volume change is introduced and variables storing initial starting volumes are updated. Cells (TCs and DCs) are represented by agents that store information about interaction history and present state. The ‘Context’ describes the environment and ‘Projections' between agents allow information transfer. Each
time-step represents 20s and model outputs are recorded every, 1, 15 or 180 time-steps. B) Following equilibration, the ‘main’ function calls a repeated series of sub-functions that describe DC arrival and TC response. Each function contains further actions, with descriptions available within the supplementary (S2 File).

Figure 4: Model visualization. A) The environment is divided into grid compartments and in this case, the TCs are color-coded based on time spent in the paracortex. B) The grid (cubes with sides of 6μm) is used to distinguish the neighboring areas of cells and cell migration.

Calibration and Validation

Baseline simulations with limited paracortical swelling (V_{max}=1.2) were performed to ensure the baseline parameters (listed in S1 File) produced realistic TC response dynamics. Following calibration, we carried out a series of four simulated experiments varying a single parameter at a time to mimic in-vivo and/or in-vitro experiments varying the corresponding biological parameter. First, we inhibited the downregulation of S1P\(_1\)r on activated TCs in-silico and then compared the effects on the total activated and cognate CD4\(^+\) and CD8\(^+\) TCs present to the effects observed during corresponding in-vivo experiments by Lo \textit{et al} and Gräler \textit{et al} [33,73]. Second, we varied the initial proportion of cognate TCs in-silico between 1.0x10\(^{-5}\) and 1.4x10\(^{-4}\) and compared the resulting CD4\(^+\) and CD8\(^+\) effector TC response to the relationship between cognition and effector TC number observed in in-vivo and in-vitro by Obar \textit{et al} and Moon \textit{et al} [72,74]. In the third simulation set, we varied the number of agDCs as a fraction of total TCs (Φ_{DC}) incrementally from 0.04 to 0.0015 (2200 to 77 agDCs). We then compared the CD8\(^+\) response in-silico to the reported CD8\(^+\) effector TC
response found in-vivo and in-vitro by Kaech et al, who injecting mice with varying bacterial antigen dose or agDCs, and by Martín-Fontecha et al who applied antigen-pulsed DCs to TC culture [75,76]. Fourth, we simulated early DC apoptosis in-silico and compared the effect on the total number of CD8+ effector to the results of Prlic et al, whereby agDC lifespan was curtailed in rats by injecting diptheria toxin (DT)-sensitive agDCs into the LNs, then eliminating the DCs with DT injection, prior to natural DC apoptosis [77].

Varying lymph node swelling and TC transit

TC response and changes in TC and DC contacts were analysed while varying maximal fold-increase in swelling volume (VMax) from 1 to 2.8, and while varying the required number of TCs present to reach half of VMax (Tmid, Figure 2A) between 8x10^4 and 13x10^4. We further investigated the impact of TC retention by modulating S1P1r expression on newly differentiated TCs (effector TCs having undergone 4-6 proliferations) and varying the initial inflammation-triggered naïve TC retention. To assess the robustness of our primary findings, alternative models were then considered. We compared TC response and changes in TC and DC contacts when modelling paracortical volume as directly proportional to TCs present. We also assessed changes in effector TC response when TCs were permitted to re-enter HEVs, or increased exit probability, if two TCs occupied an exit compartment, and when exit area diameter remained a fixed proportion of the paracortical radius. We used application of ANOVA to detect variance and quantified correlation by calculating Pearson’s coefficient, and the R² value. A significance threshold of p<0.05 was used throughout.

Global sensitivity analysis


To identify any influential but uncertain or biologically unconstrained parameters, we carried out a global sensitivity analysis. A total of 300 parameter combinations were selected using Latin Hypercube sampling to reduce sample size while ensuring inclusion of samples near the minimum and maximum parameter ranges. Simulations using each combination were repeated 3 times. Partial-rank correlation coefficients (PRCCs) between each parameter and output of interest (activated TCs, effector TCs, memory TCs, effector TCs exited and memory TCs exited) were calculated for each day (3-13). This was to quantify the effects of a single parameter while accounting for the effects of other parameters, assuming monotonic relationships [78]. Significance tests were performed between the output PRCCs and zero, meaning weaker correlations may be still be significant [78]. We report all significant PRCCs with strength greater than 0.2 (see S4 File).

Results

The model produces realistic baseline TC motility and response to agDCs

By tracking and recording the TC movement (Fig 5A, B), we confirmed that the calibrated model produced an average TC velocity (n=200) of 13.1 μm/min, reaching up to 24 μm/min (Fig 5C), in-line with murine in-vivo measurements [41,43,56,79,80]. The linear relationship between TC displacement versus square root of time, (Fig 5D, E) illustrated that random walk behaviour was maintained [81]. The motility coefficient (CM), analogous to a diffusion coefficient, was 63.2 μm²/min, which is within the range of 50-100 μm²/min observed in mice [38]. The mean TC paracortex transit time was 13.1 hours (n=16,000), ranging from 20 minutes to >60 hours (Fig 5F), in-line with observations that 74% of CD4⁺ TCs and 64% of CD8⁺ TCs transit murine LNs within a day [82].
Figure 5. Baseline TC motility. A) TC tracking through the paracortex with individual tracks displayed. TCs entered centrally, representing entry from HEVs, and at the top of the paracortex, representing entry in afferent lymph (yellow arrow). TCs exited at the periphery (green arrows). B) Tracks of 200 TCs (each colour represents one TC path) transposed to the same origin point display the random walk behaviour outwards from the origin. C) Most TCs completed transit in <24hrs. D) Mean TC velocity over the entirety of their path. E) Mean and SEM (n=100) of TC displacement from entry point. F) Mean (+/−SEM) of TC displacement showed a linear relationship to the square root of time, indicating that random walk behaviour is maintained.

TC responses to agDC stimuli corresponded well to experimental data from cell-culture models and in-vivo trials in mice, sheep and rats, displaying the expected phases of TC trafficking and response (Fig A in S3 File). TC numbers began to increase approximately 6 hours after initial entry of agDCs, and by day 11 had returned to within 15% of pre-stimulus values (Fig 6A), in-line with temporal responses observed in-vivo [6,7,15]. The appearance of activated, effector, and memory TCs began at 16-24 hours, day 3.5 and day 5 post-agDC entry, respectively, in agreement with cell-culture models and in-vivo observations [83,84]. Effector CD4+ TCs appeared 1-1.5 hours before CD8+ effector TCs (Fig 6A, B). As observed in-vivo, the peak population of cognate CD8+ TCs was an order of magnitude higher than that of CD4+ TCs (Fig 6C) [69,85]. The contraction phase began at day 7 and continued through day 11. The increase in TC egress rate peaked a day later than the increase in TC entry rate (Fig 6D), corresponding well with in-vivo observations of delayed increase in TC egress and TC recruitment dynamics [11,86].
Figure 6. TC responses in the paracortex following entry of agDCs under baseline conditions. Average result with SEM of 12 simulations. Note the differing vertical scales.

Incoming agDCs numbers are depicted by the dot-dashed line in B (right axis). The total number of TCs (black line in A) peaked in the paracortex at 3.5 days and was comprised mainly of non-cognate naïve TCs, as the number of effector TCs (blue line A) peaked at day 6. The appearance of activated TCs (red line in B) began 12 hours after the first agDCs entered. Memory TCs (black line in B) began to appear at 5 days and 25% of the peak number of memory TCs remained at the end of the simulation. Cognate CD4+ TCs (Green line in C) began to proliferate extensively at day 2.2. Cognate CD8+ TCs (purple line C) began to proliferate at day 4 and reached numbers 10x more than cognate CD4+ TCs. D) TC entry rate increased 2x, peaking at day 3, whilst TC egress rate declined between day 1 and 2, then increased 3x by day 4.

Following calibration, the model was validated through replication of in-vivo and in-vitro experiments

Whilst holding the calibrated parameters, in each of four simulated experiments, a single parameter was varied to mimic published in-vivo and in-vitro experiments. Abrogating S1P1r down-regulation after antigenic stimulus detection in-silico reduced the number of activated TCs in the paracortex by 60%, 72% and 81% at $V_{\text{max}}$=1.2, 2.0 and 2.5, respectively (Fig 7D). This was a smaller reduction than observed during in-vivo experiments when activated TCs maintaining S1P1r expression were transferred to LNs, resulting in 90% less activated TC retention in the LNs 15 hours later compared to control mice (Fig 7A) [33]. The in-silico reduction was, however, greater than the 40% reduction in activated TCs found with constitutive TC expression of S1P1r in-vivo post-immunisation (Fig 7B) [73]. The in-silico
The total number of CD4⁺ and CD8⁺ effector TCs reached 20% and 6% respectively of the control response when S1P₁r was abrogated (Fig 7E, F). This is similar to the 27% and 5% of control response recorded with constitutive inhibition of S1P₁r expression in-vivo (Fig 7C) [73]. See Fig B,C & D in S3 File for three additional sets of validation results.

**Figure 7. Comparison between in vivo assessments of the effects of S1P₁r downregulation and model predictions.** Top row contains experimental data, and bottom row shows comparable modelling results. A. Pre-activated wildtype TCs (S1P₁r) and TCs over-expressing S1P₁r (S1P₁r++) were transferred into mice and further TC entry was blocked. After 15 hours there was a 90% reduction in retention of S1P₁r++ activated TCs. Adapted from Lo et al 2005. B) Activated TC number in the LNs 24 hrs post-transfer dropped by 40% in transgenic mice with constitutive S1P₁r expression while (C) proliferative CD4⁺ and CD8⁺ TC response decreased to 20% & 6% of that of the wild-type mice. Adapted from Gräler et al 1997. D-F). The results of simulations (n=10) preventing S1P₁r downregulation (-SP regulation) compared to baseline simulations (+SP regulation) showed similar trends to these in-vivo results (which were not used in model construction). Mean (+-SEM) total activated TCs number was reduced 60%, 72% and 81% at V_max=1.2,1.5 & 2. E) The mean (+-SEM) total CD8⁺ TC number, was diminished to 25,15 and 18% of control response at V_max=1.2,1.5 & 2. F) CD4⁺ TC number was similarly diminished to 8-10% of control at all V_max values.

The results of the global parameter sensitivity analysis indicated that the dominant parameters in determining the target outcomes of TC activation, total TC effectors, and TCs exited were F_cog, T_DCin, V_max. The unconstrained parameters used to describe signal integration, such as mean amount of signal accumulated to activate or differentiate, were not identified as
significantly influential (p>0.05 and R^2>0.2) in determining the target outcomes (see Fig A and Tables A-C in S4 File).

**Paracortical swelling consistently aids TC activation but not effector TC production**

When maximal swelling (V_{max}) was varied from 1 to 2.8, activated TC number positively correlated with V_{max} (R^2= 0.96, p<10^{-5}) doubling in number (Fig 8A). However, the total number of effector TCs negatively correlated with V_{max} (R^2=0.86, p<10^{-3}) and decreased by 15% (Fig 8B). Neither the number of effector TCs that exited by day 10, nor total number of cognate CD4^+ TC present, varied significantly (Fig E.ii and E.iii in S3 File). However, total cognate CD4^+ TCs that left the paracortex by day 10 increased by 30% and positively correlated with V_{max} (R^2=0.76, p=0.001) (Fig 8C). Conversely, there was no change in the number of exiting cognate CD8^+ TCs with increasing V_{max} (Fig E.iv in S3 File), but the total number of cognate CD8^+ TCs present decreased by 25% and negatively correlated with V_{max} (R^2=0.855, p<10^{-3}) (Fig 8D). The peak rate of TC recruitment positively correlated with V_{max}, and TC egress rate increased with V_{max} from day 3-6 (Fig 8E, F). The pattern of increased TC activation but decreased effector TC production remained when LN volume increased as a linear function of TCs (Fig A in S5 File).

**Figure 8. The variation in TC subsets as V_{max} was varied from 1 to 2.8 with T_{mid}=10^5.**

A) The total activated number of TCs in the paracortex incrementally increased with V_{max} and doubled in number between V_{max}=1 and 2.8. (R^2 = 0.96, p=1.07x10^{-6}). B) The total number of effector TCs decreased 0.3x with increasing V_{max} (R^2=0.86, p=1.23x10^{-4}). C) The total number of cognate CD4^+ TCs that exited positively correlated with V_{max} (R^2=0.76, p=0.001) and increased 1.3x. D) The total cognate CD8^+ TCs negatively correlated with V_{max}
Peak entry and peak exit rate increased proportionally to $V_{\text{max}}$. All results are the mean of $n \geq 7$ simulations with SEM displayed.

**Varying the ease of swelling influences resulting TC populations**

In some cases, varying the required number of TCs present to reach half the maximal swelling ($T_{\text{mid}}$) from a default value of $10^5$ counteracted the effect of varying maximal swelling of LNs on effector TC production. Simulations were carried out varying $T_{\text{mid}}$ from the default of $10^5$ with a lower $T_{\text{mid}}$ of $8 \times 10^4$ (leading to swelling beginning a day earlier (Fig 9A) or higher $T_{\text{mid}}$ of $13 \times 10^4$ TCs, alongside a high or low maximal swelling ($V_{\text{max}}=1.2$ or 2.5). We found that, regardless of $T_{\text{mid}}$, at least 40% more activated TCs were recorded when $V_{\text{max}}$ was high compared to when $V_{\text{max}}$ was low (Fig 9B). At a low $T_{\text{mid}}$, when swelling was increased, there was a significant drop in effector TCs number (Fig 9C). However, at a high $T_{\text{mid}}$, increasing $V_{\text{max}}$ no longer negatively impacted effector TC exit. This change in effector TC response was a result of a variation in typical cognate CD8$^+$ TC response. As shown in Fig 8D previously, cognate CD8$^+$ TC number decreased with swelling; this was similarly observed in Fig 9D at a lower $T_{\text{mid}}$ of $8 \times 10^4$. However, at a higher $T_{\text{mid}}$, cognate CD8$^+$ TC numbers increased with swelling, whereas there was no change associated with $T_{\text{mid}}$ in cognate CD4$^+$ TC response (Fig 9E).

When $V_{\text{max}}$ was varied in 0.2 increments with a higher $T_{\text{mid}}$ of $12 \times 10^4$, the number of effector TCs no longer decreased with $V_{\text{max}}$ (Fig 9F). An intermediate swelling of 1.8-fold was identified as the optimum $V_{\text{max}}$ to produce effector TCs with 7% more total effector TCs recorded than at any other value of $V_{\text{max}}$. Varying both maximal swelling and $T_{\text{mid}}$ over a wider range confirmed that swelling size and early swelling increase TC activation, with $T_{\text{mid}}$
correlating negatively with activated TCs at every value of $V_{\text{max}}$ ($R=0.79-0.99$) (Fig E.viii in S3 File). The effect of positive correlation of $T_{\text{mid}}$ with effector TCs exited is only clear at larger swelling ($V_{\text{max}}=2.5$) (Fig 9G), which is likely due to the more significant impact of varying $T_{\text{mid}}$ with larger swelling (Fig 9A).

**Figure 9. Increasing $T_{\text{mid}}$ counteracts the loss in effector TCs with $V_{\text{max}}$.** Swelling was varied with $T_{\text{mid}}=8\times10^4$ and $T_{\text{mid}}=1.3\times10^5$ instead of default (Figure 8). A) The paracortex swells earlier and for a longer duration with a low $T_{\text{mid}}$. B) TC activation increased with larger swelling, but a higher $T_{\text{mid}}$ also reduced activation at a large swelling ($p=0.002$). C) With a lower $T_{\text{mid}}$, as swelling increased, effector TC number decreased. This effect was due to a change in CD8$^+$ TC number (D) as CD4$^+$ TC number (E) increased with $V_{\text{max}}$, but was unaffected by varying $T_{\text{mid}}$. F) The results of simulations varying $V_{\text{max}}$ in 0.2 increments with a higher $T_{\text{mid}}$ of $1.2\times10^5$ instead of baseline $10^5$, eliminated the negative correlation between effector TCs and swelling. (G) Further simulations varying $T_{\text{mid}}$ confirmed that at a large swelling ($V_{\text{max}}=2.5$), delayed swelling with a higher $T_{\text{mid}}$ resulted in more total effectors TCs ($R^2=0.97, p=3\times10^{-4}$).

**Analysis of underlying TC and DC contacts did not identify altered TC and DC contacts as an influential driving mechanism.**

We next wanted to better understand whether changes in TC and agDC interactions may be a driving mechanism for changes in effector TC production. When analysing the mean number of interactions with DCs by cognate TCs present at 3 days, there was no trend with swelling (Fig 10A), even when considering the time since the TC entered (Fig 10B). We then instead looked at the mean number of cognate TCs that DCs interacted with over their lifespan. Correlation between swelling and interactions became more significant the later the DCs
entered, beginning from quartile 2, representing DCs that enter at approximately 15-30 hours, (R²=0.69, p=0.002) to quartile 4 (R²=0.83, p=2.3x10⁻⁴) (Fig 10D). When T_mid was increased to 12x10⁵ TCs, there was no significant correlation between TC and DC interaction and V_max (Fig 10E). When T_mid was varied between 8x10⁴ and 12x10⁴, no significant correlation with the number of interactions with DCs by cognate TCs was found at any swelling value (Fig 10C), while the number of cognate TCs contacted by DCs showed a positive correlation for DCs in quartile 4 with a large maximal swelling of V_max =2.5 only (R²=0.72, p=0.03) (Figure 10F).

Taken together, these results suggest that the availability of DCs to the cognate TCs is not significantly influenced by the swelling at day 3, a timepoint that corresponds with peak swelling. When swelling was varied and T_mid fixed, the DCs that entered during quartile 3 contacted less cognate TCs as V_max increased (Fig 10D), but the cognate TCs present did not contact any less DCs at day 3 (Fig 10A). Although during peak swelling time there isn’t a clear relationship between contacts and TC response, our model is not constructed such that we can ascertain whether changes in the number of cognate TCs that DCs contact during quartile 4 (day 4-6) are a result of, or contribute to, the increase or decrease in cognate TCs present.

**Figure 10. Changes in cognate TCs and agDC interaction with varying swelling.** The mean number of interactions with agDCs recorded from cognate TCs present at day 3; (A) showed no trend with swelling, (B) showed no trend with swelling when taking into account the time spent by the TC in the paracortex and (C) showed no trend with T_mid when swelling was held constant. The mean number of cognate TCs that a DC interacted with during its lifespan was recorded, and the DCs grouped dependent on entry time (first 25% (Quartile 1) etc. to enter). With default T_mid of 10⁵ (D), a very weak negative trend with swelling occurred
during Quartile 1 ($R^2=0.48$, $p=0.05$), that became stronger the later the DC entered ($R^2=0.69$, $p=0.002$; $R^2=0.77$, $p=6.9\times10^{-4}$; $R^2=0.83$, $p=2.3\times10^{-4}$ for quartile 2, 3 and 4 respectively). (E) With a higher $T_{mid}$ of $1.2\times10^5$, no clear trend with swelling was identified. F) Only at a large swelling did DCs that entered later (Quartile 4) contact more cognate TCs when increasing $T_{mid}$ ($R^2=0.72$, $p=0.03$).

S1P1r-mediated temporary retention of early effector TCs increased TC response.

Regardless of permitted maximal swelling, down-regulating S1P1r on early effector TCs to less than 80% of naïve TC expression ($SP_{early}<0.8$) produced a sustained increase in total TCs, despite the action only directly affecting the small subset of early effector TCs (Fig 11A). We also found that increasing the S1P1r downregulation by lowering $SP_{early}$ from its default value of 0.4 resulted in changes in effector TC number that far outweighed the changes due to swelling (Fig 11B). $SP_{early}$ inversely correlated with number of effector TCs at every swelling value ($R^2=0.92, 0.93, 0.92$, $p<0.005$). Reducing $SP_{early}$ from 0.4 to 0.05 doubled the number of effectors TCs exiting, while increasing $SP_{early}$ to 0.8 at least halved the number (Fig 11B). When analysing the TC sub-populations, both CD4+ and CD8+ effector TCs that exited the paracortex by day 10 negatively correlated with $SP_{early}$, and doubled with $SP_{early}$ from 0.4 to 0.05 (Fig 11C, D). This indicates that CD4+ TCs do maintain proliferative capacity under the correct conditions.

Interestingly, the number of TCs contacted by DCs increased with decreasing $SP_{early}$, but overall decreased with swelling (Fig 11E). Unlike in our previous simulations permitting swelling with default fixed $SP_{early}$ (Fig 10), the effector TC number did not decrease with swelling, suggesting that increased contact is not a driving mechanism. We also took note of the changes in the TC to grid ratio while varying $SP_{early}$ (Fig 11F). With a small $V_{max}$,
increased TC crowding was observed as expected. However, this retention alone did not
produce the highest number of effector TCs. Instead, the combination of retention and
swelling produced approximately 15% and 10% more effector TCs with V_max of 2.0 and 2.5
respectively, outweighing any negative effects of increased TC egress that accompany
swelling (Fig 11G). Implementation of an alternative model with non-specific constraint of
TC egress by reducing the expansion in the cap-shaped exit area also resulted in increased
effector TC exit but prolonged swelling above a 1.4x swelling (Fig B in S5 File).

**Figure 11. Temporary retention of effector TCs by modulating S1P_1r expression on
newly differentiated TCs (SP_early).** A) Reducing SP_early resulted in a higher total number of
TCs in the paracortex. B) Effector TCs that exited the paracortex were influenced more by
SP_early than by V_max, negatively correlating with SP_early (R^2>=0.92, p<0.005). SP_early
negatively correlated with both (C) CD8^+ TCs exited (R^2>0.92, p<0.005) and (D) CD4^+ TCs
exited (R^2>0.91, p<0.005). E) The mean number of cognate TCs contacted increased as
SP_early was lowered to 0.1 at each value of V_max but overall decreased with V_max. F) At a
smaller maximal swelling, a higher TC to grid ratio was maintained but this did not translate
to more or less effector TC production. (G) Reduced SP_early resulted in higher overall rates of
TC egress in week 2 of the simulation.

Non-specific early LN shutdown with a doubling of LN volume permitted did not
significantly impact effector TC production

We also varied the degree of initial LN shutdown (whilst permitting a doubling of LN
volume) by downregulating SP_inflam from 0.1 (90% downregulation) to SP_inflam=1 (no
shutdown). Such downregulation could occur in vivo in response to inflammation. As
SP_inflam decreased, TC activation increased (R^2= 0.83, p=0.01) (Fig 12B), but varying LN
shutdown showed no trend on effector TCs produced (Fig 12C). No correlation between mean number of contacts with DCs by cognate TCs present at day 3 and increased LN-shutdown was observed, but a positive correlation with number of DCs contacted by non-cognate TCs was found ($R^2=0.93$, $p=0.0017$) (Fig 12D). Increasing $S1P_1r$ down-regulation from 60 to 90% resulted in a sharp, 3-fold higher peak in the total number of TCs (Fig 12A) and paracortical volume, which would not be expected physiologically.

Figure 12 Varying LN shutdown by modulating inflammation-induced $S1P_1r$

downregulation ($SP_{inflam}$). Modulation of $S1P_1r$ down-regulation in response to initial detection of inflammatory signals from no downregulation ($SP_{inflam}=1$) to 90\% downregulation ($SP_{inflam}=0.1$), A) total TC number decreases several fold but (B) TC activation increases ($R^2= 0.83$, $p=0.01$) while (C) Effector TC production shows no trend. D) Non-cognate TCs contact more DCs as $SP_{inflam}$ is increased ($R^2=0.93$, $p=1.7\times10^{-3}$).

Discussion

In this work, we aimed to better understand the role of lymph node swelling and other adaptive immune processes in the formation of TC responses. Our study builds on previous studies that used ABMs to investigate the impact of stimuli, interaction dynamics, signal integration kinetics and TC migration on effector and memory cell generation in the paracortex, with the aim of better understanding how to manipulate immune response [44-46,48,50]. We focused our results analysis on macroscale alterations and accompanying changes in egress and recruitment. The model was calibrated to baseline conditions that produce a TC response comparable to one set of in-vivo results and validated by confirming the in-silico TC response matched those observed in a combination of separate in-vivo and in-vitro studies. By varying paracortical swelling and $S1P_1r$ expression, we showed that
paracortical swelling aids TC activation, but early swelling can impair effector TC response.

However, temporary retention of newly differentiated TCs influences the overall effector TC response more than swelling, providing a potential mechanism to overcome swelling-related impairment. Analysis of the underlying mechanisms suggest that the changes in egress probability are more influential than alteration in TC and DC contacts.

A key finding was that, although paracortical swelling facilitated activated TC production over a 5-day period, subsequent effector TC numbers only increased with additional non-specific exit constraints or specific early effector TC retention. During swelling, TC trafficking rate is increased with swelling while activated TCs are retained (Fig 8). Increased TC recruitment from HEV implies that more cognate TCs enter, facilitating screening and leading to more TC activation, an effect that is amplified as the activated TCs proliferate [9].

However, we found that increased activation with larger and/or earlier swelling did not consistently lead to more effector TCs. Instead, swelling that occurred too early reduced the total number of effector TCs and the number of effector CD8+ TCs that exited, and without additional S1P1r retention or constrained exit, swelling only ever reduced or maintained effector TC production (Fig 9, S4 File). Our results also resemble those from in-vivo experiments showing that TC activation is depleted with removal of S1P1r-mediated retention, but also that under these conditions, swelling also no longer aids TC activation (Fig 7).

We found that although swelling is well underway by day 3, there was no clear impact on the mean number of interactions with agDCs by cognate TCs (Fig 10). We also found that by day 4.5, the number of cognate TCs that each agDC has interacted with decreased with swelling. Whether this is due to, or contributes to, less cognate TCs present is unclear. By day 6, most
DCs have undergone apoptosis. When alternative models of egress were considered, increased TC activation with swelling could result in increased effector TC production, when the exit area growth with swelling was constrained (Fig B in S5 File). It thus appears that delayed swelling allows for TC recruitment and retention to be a stronger initial influence than TC egress, by reducing access to exit points and thus the egress rate.

Another key finding was that temporary S1P1r-mediated retention of newly differentiated effector TCs could counteract effects due to swelling that resulted in reduced effector TC number (Fig 11). As S1P1r downregulation decreased, TC and DC contact increased. With swelling, TC and agDC contact decreased but the numbers of exiting effector TC remained constant. This further suggests increased retention has a stronger effect than increased agDC contact. Non-specific retention in the first few days had no impact on effector TC response (Fig 12).

The temporary nature of this modulation at a later timepoint is crucial to increase effector TC number. Therapeutically, down-regulation of TC S1P1r is currently the mechanism of multiple sclerosis drug Fingolipid to maintain effector TCs in the LN indefinitely [87]. This is to prevent autoimmune response in the brain but also results in reduced lymphocyte numbers in the blood due to a block of TC egress from the thymus. Permanent inhibition of S1P1r expression on effector TCs only has also been carried out in-vivo [25]. However, temporary downregulation on selectively newly differentiated TCs may prove technically difficult, suggesting identification of alternative means of retention is desirable.

We also found that early swelling hindered CD8+ effector TC response more than that of CD4+ TCs (Fig 9). However, increased S1P1r downregulation by early effector TCs also
increased CD4+ TC production, indicating that the capacity for further CD4+ expansion remained. The earlier and shorter duration of proliferation of CD4+ TCs may mean that further proliferation is proportionally less affected by increases in TC egress that accompany paracortical swelling compared to CD8+ TCs [69].

The insight provided by our results imply chronically inflamed LNs could be addressed by modulating newly-differentiated effector TC egress with swelling through S1P1r expression, or by manipulating access to exit areas. Physiologically, factors that could be manipulated to alter the effective T_{mid} include stromal cell contraction and proliferation rate, the sensitivity of immune response to antigen signalling, and the sensitivity to accumulated fluid and TCs. Astarita et al. 2015 induced swelling by transferring 10^6 cognate TCs directly into murine LNs, whilst also effectively modulating T_{mid} by inducing FRC elongation and inhibiting FRC contraction by manipulating the CLEC-2-podoplanin axis [12]. In contrast to our predictions, subsequent TC proliferative response was enhanced by the facilitated swelling. However, as both in-silico and in-vivo experiments have shown, proliferative response size is proportional to the starting frequency of cognate TCs, an effect that we also observed in our model [22,72,74] (Fig B in S3 File). We suggest that with an inflated number of initial cognate TCs, the time-point at which swelling of LNs becomes helpful is likely shifted forward. Our model could be falsified experimentally by showing that applying the minimal antigenic stimulation required to elicit an immune response will produce more effector TCs when swelling is facilitated than when swelling is impeded.

The purpose of the global sensitivity analysis was to identify parameters having strong influence on outputs of interest, with the intent of focusing future experimental efforts on their identification. This also aids in moderating conclusions influenced by unidentified
parameters. For the majority of the model parameters, where a relevant biological basis exists, we parameterised the model using data from *in-vivo* studies. For parameters \( \text{MHCI}_{1/2} \) and \( \text{MHCII}_{1/2} \) (respective MHC half-lives), estimations were based on *in-vitro* studies but had little influence on our outputs of interest in comparison to other parameters. Unconstrained parameters in the model relate to the signal gain and loss (\( K_s \) and \( \gamma \)), the characterization of activation and differentiation probabilities (\( \text{Act}_{\mu 4^+} \), \( \text{Act}_{\mu 8^+} \), \( \text{Diff}_{\mu 4^+} \), \( \text{Diff}_{\mu 8^+} \)), TC recruitment and paracortical swelling (S4 File). The sensitivity analysis showed that the parameters populating the activation and probability curves were not highly influential. The influence of signal integration patterns has been the focus of previous modelling studies [46,50,66].

Model robustness was assessed by examining whether the model could replicate a variety of *in-vivo* and/or *in-vitro* experiments. Most scenarios of varying stimuli strength, duration and TC cognition rate correlated well with such experiments. In some cases, slight differences were observed, for example, the effect of *in-silico* elimination of agDCs after 12 hours more closely matched *in-vivo* elimination of DCs after 1 hour (Fig C in S3 File). This is possibly because, although the majority of DCs are eliminated rapidly post-DT injection, complete removal of DCs *in-vivo* required additional time (12 hours) [75]. Prevention of S1P1r downregulation on activated TCs *in-silico* resulted in a smaller reduction in activated TC number (60-80%) than the 90% described by Lo *et al.* 2004, but more than the 40% reported by Gräler *et al.* 1997 (Fig 7) [33,73].

Our model captures both macroscale and microscale interactions, but limitations remain, including the use of a simplified LN geometry, lack of chemotactic influences, and no direct representation of the FRC network. TCs display a random walk regardless of FRC network presence, and previous models suggest TC contacts are not significantly influenced by its
The fidelity of the model is also limited by a lack of information on the effects of swelling on the availability of medullary exit points. TCs may enter the medulla via specialised junctions at specific fixed ‘hotspots’ or at cell junctions throughout the tissue [35]. Our model has identified this as an important area for future investigation. We omitted DC migration, but our results indicate that DC availability is not a limiting factor.

Lack of a precise mechanism to drive LN swelling is also a challenge, but use of a sigmoidal function to dictate paracortical volume captured the typical phases of swelling better than alternative models (S4 File). An initial TC increase is permitted without triggering significant swelling, reflecting initial inhibition of stromal cell proliferation by increased secretion of IFN type 1 [88]. The delayed increase in volume in response to TC number then represents the switch in signalling at day 2 to favour LEC proliferation and expansion of LNs, through mechanisms such as DC-induced secretion of VEGF from stromal cells and increased elasticity of FRC network [6,89]. We have also neglected the role of chemokines in retaining TCs [18]. When both CCR7 and S1P1r expression on TCs is inhibited in-vivo, TCs migrate to the edges of the paracortex but cannot exit due to lack of S1P1r expression [25]. Accordingly, inclusion of S1P1r down-regulation was prioritised over CCR7, but the strong influence of retention suggests future models should include a wider range of retentive influences [58].

In future model iterations, inclusion of additional factors, such as lymph flow and pressure alterations (along with fluid exchange with nodal blood vessels), could also significantly improve the representation of swelling, and thus TC egress and retention. It has been well-established that changes in hydrostatic and oncotic pressure differences across nodal blood vessel walls can reverse the net fluid exchange [90,91]. Afferent lymphatic flow to the LNs also increases with immune response. Therefore, a key next step is to couple the ABM to a
computational flow model, with resident DCs in the LNs also added, and the maximal permitted swelling increased to confirm the observed trends. Lymph flow also influences chemokine concentration fields, and likely mechanoresponsive cell expression of signalling molecules and receptors.

**Conclusion**

Our results suggest that, although swelling of LNs may aid TC activation, avoiding excessive swelling of LNs and/or temporary retention of effector TCs may boost effector TC response when initial TC response is small, for example, in immuno-suppressed patients, or desirable, such as when optimizing vaccine design to minimise antigen dose. Our model has proved a useful tool to identify the importance of better understanding alterations in TC egress with swelling. Moreover, retention of newly differentiated TCs via the modulation of the TC receptor Sphingosine-1-phosphate-1-receptor (S1P₁r) showed that selective modulation on the small subset of effector TCs could strongly impact the efficiency of TC response, and even overcome loss in efficiency of TC response due to early enlargement. Although permanent blockade of effector TC egress has been utilised to treat multiple sclerosis, temporary retention of effector TCs to boost subsequent effector TC production presents as a novel mechanism. This finding also emphasizes the influence that retentive features, including factors such as chemokines, may have on effector TC response, which may be more practical in-vivo targets to manipulate.

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Supporting Information

S1 File. Supplementary Methods A. Further description of modelling methods. 1.1 TC recruitment, 1.2 Agents and agent migration, 1.3 Agent interaction and signal integration. Fig A & B, and parameter Tables A & B.

S2 File. Supplementary Methods B. UML diagrams.
S3 File. Supplementary Results A. Supplementary model calibration (Fig A), model validation (Fig B-D) and LN swelling results (Fig E).

S4 File. Supplementary Results B. Global sensitivity analysis results. Fig A & Tables A-C.
The sensitivity of output activated, effector and memory TCs present and effector and memory TCs exited to parameters over the time course of the simulation.

S5 File. Supplementary Results C. Alternative models. 5.1 & Fig A. Consideration of an alternative model with linear TC to LN volume relationship. 5.2 & Fig B. Considering alternative models of TC crowding and egress.
Fig 1. Model geometry and structure. (A) LN structure displaying the pathway of arriving lymphatic fluid containing agDCs. (B) The model contains a spherical paracortex. TCs enter in the centre and exit near the interface with the medulla and SCS. The paracortex radius expands as a function of TCs present. (C) At the microscale, TCs move to adjacent grid-compartments (6 of 8 possible movements shown), interact with neighbouring agents and are influenced by the grid-compartment properties, which are updated each time-step. Microscale signals can affect TC recruitment, TC exit and stimulation via interaction with DCs. (D) Interactions between micro and macro scales.
Figure 2. Modelling the changes in paracortical volume, TC recruitment and TC expression of S1P1r. A) Previously published experimental data from studies of multiple influences on LN swelling were used to inform the selection and parameterisation of a sigmoidal function to drive LN swelling based on TC numbers. B) TC recruitment is driven by initial antigenic signal (eq. 2), with parameterization based on in-vivo data, and a proportional increase due to expansion of HEVs. C) S1P1r expression mediated the retention of (i) naïve TCs post-transmigration from the blood, (ii) naïve TCs during LN shutdown and (iii) activated and early effector TCs.
Figure 3: Structure of the code to set up and run the model. A) The model is initiated in the absence of stimulus, then capacity for paracortical volume change is introduced and variables storing initial starting volumes are updated. Cells (TCs and DCs) are represented by agents that store information about interaction history and present state. The ‘Context’ describes the environment and ‘Projections’ between agents allow information transfer. Each time-step represents 20s and model outputs are recorded every, 1, 15 or 180 time-steps. B) Following equilibration, the ‘main’ function calls a repeated series of sub-functions that describe DC arrival and TC response. Each function contains further actions, with descriptions available within the supplementary (S3 File).
Figure 4: Model visualization. A) The environment is divided into grid compartments and in this case, the TCs are color-coded based on time spent in the paracortex. B) The grid (cubes with sides of 6μm) is used to distinguish the neighboring areas of cells and cell migration.
Figure 5. Baseline TC motility. A) TC tracking through the paracortex with individual tracks displayed. TCs entered centrally, representing entry from HEVs, and at the top of the paracortex, representing entry in afferent lymph (yellow arrow). TCs exited at the periphery (green arrows). B) Tracks of 200 TCs (each colour represents one TC path) transposed to the same origin point display the random walk behaviour outwards from the origin. C) Most TCs completed transit in <24hrs. D) Mean TC velocity over the entirety of their path. E) Mean and SEM (n=100) of TC displacement from entry point. F) Mean (+/-SEM) of TC displacement showed a linear relationship to the square root of time, indicating that random walk behaviour is maintained.
Figure 6. TC responses in the paracortex following entry of agDCs under baseline conditions. Average result with SEM of 12 simulations. Note the differing vertical scales. Incoming agDCs numbers are depicted by the dot-dashed line in B (right axis). The total number of TCs (black line in A) peaked in the paracortex at 3.5 days and was comprised mainly of non-cognate naïve TCs, as the number of effector TCs (blue line A) peaked at day 6. The appearance of activated TCs (red line in B) began 12 hours after the first agDCs entered. Memory TCs (black line in B) began to appear at 5 days and 25% of the peak number of memory TCs remained at the end of the simulation. Cognate CD4$^+$ TCs (Green line in C) began to proliferate extensively at day 2.2. Cognate CD8$^+$ TCs (purple line C) began to proliferate at day 4 and reached numbers 10x more than cognate CD4$^+$ TCs. D) TC entry rate increased 2x, peaking at day 3, whilst TC egress rate declined between day 1 and 2, then increased 3x by day 4.
Figure 7. Comparison between in vivo assessments of the effects of S1P1r downregulation and model predictions. Top row contains experimental data, and bottom row shows comparable modelling results. A. Pre-activated wildtype TCs (S1P1r) and TCs over-expressing S1P1r (S1P1r++) were transferred into mice and further TC entry was blocked. After 15 hours there was a 90% reduction in retention of S1P1r++ activated TCs. Adapted from Lo et al 2005. B) Activated TC number in the LNs 24 hrs post-transfer dropped by 40% in transgenic mice with constitutive S1P1r expression while (C) proliferative CD4+ and CD8+ TC response decreased to 20% & 6% of that of the wild-type mice. Adapted from Gräler et al 1997. D-F). The results of simulations (n=10) preventing S1P1r downregulation (-SP regulation) compared to baseline simulations (+SP regulation) showed similar trends to these in-vivo results (which were not used in model construction). Mean (+-SEM) total activated TCs number was reduced 60%, 72% and 81% at V_max=1.2, 1.5 & 2. E) The mean (+-SEM) total CD8+ TC number, was diminished to 25,15 and 18% of control response at V_max=1.2, 1.5 & 2. F) CD4+ TC number was similarly diminished to 8-10% of control at all V_max values.
Figure 8. The variation in TC subsets as $V_{\text{max}}$ was varied from 1 to 2.8 with $T_{\text{mid}}=10^5$. A) The total activated number of TCs in the paracortex incrementally increased with $V_{\text{max}}$ and doubled in number between $V_{\text{max}}=1$ and 2.8. ($R^2=0.96$, $p=1.07\times10^{-6}$). B) The total number of effector TCs decreased 0.3x with increasing $V_{\text{max}}$ ($R^2=0.86$, $p=1.23\times10^{-4}$). C) The total number of cognate CD4$^+$ TCs that exited positively correlated with $V_{\text{max}}$ ($R^2=0.76$, $p=0.001$) and increased 1.3x. D) The total cognate CD8$^+$ TCs negatively correlated with $V_{\text{max}}$ ($R^2=0.855$, $p=1.28\times10^{-4}$). E,F) Peak entry and peak exit rate increased proportionally to $V_{\text{max}}$. All results are the mean of $n\geq7$ simulations with SEM displayed.
Figure 9. Increasing $T_{mid}$ counteracts the loss in effector TCs with $V_{max}$. Swelling was varied with $T_{mid}=8 \times 10^4$ and $T_{mid}=1.3 \times 10^5$ instead of default (Figure 8). A) The paracortex swells earlier and for a longer duration with a low $T_{mid}$. B) TC activation increased with larger swelling, but a higher $T_{mid}$ also reduced activation at a large swelling ($p=0.002$). C) With a lower $T_{mid}$, as swelling increased, effector TC number decreased. This effect was due to a change in CD8$^+$ TC number (D) as CD4$^+$ TC number (E) increased with $V_{max}$ but was unaffected by varying $T_{mid}$. F) The results of simulations varying $V_{max}$ in 0.2 increments with a higher $T_{mid}$ of $1.2 \times 10^5$ instead of baseline $10^5$, eliminated the negative correlation between effector TCs and swelling. (G) Further simulations varying $T_{mid}$ confirmed that at a large swelling ($V_{max}=2.5$), delayed swelling with a higher $T_{mid}$ resulted in more total effectors TCs ($R^2=0.97$, $p=3 \times 10^{-4}$).
Figure 10. Changes in cognate TCs and agDC interaction with varying swelling. The mean number of interactions with agDCs recorded from cognate TCs present at day 3; (A) showed no trend with swelling, (B) showed no trend with swelling when taking into account the time spent by the TC in the paracortex and (C) showed no trend with $T_{\text{mid}}$ when swelling was held constant. The mean number of cognate TCs that a DC interacted with during its lifespan was recorded, and the DCs grouped dependent on entry time (first 25% (Quartile 1) etc. to enter). With default $T_{\text{mid}}$ of $10^5$ (D), a very weak negative trend with swelling occurred during Quartile 1 ($R^2=0.48, p=0.05$), that became stronger the later the DC entered ($R^2=0.69, p=0.002; R^2=0.77, p=6.9\times10^{-4}; R^2=0.83, p=2.3\times10^{-4}$ for quartile 2, 3 and 4 respectively). (E) With a higher $T_{\text{mid}}$ of $1.2\times10^5$, no clear trend with swelling was identified. F) Only at a large swelling did DCs that entered later (Quartile 4) contact more cognate TCs when increasing $T_{\text{mid}}$ ($R^2=0.72, p=0.03$).
Figure 11. Temporary retention of effector TCs by modulating S1P₁r expression on newly differentiated TCs (SP<sub>early</sub>). A) Reducing SP<sub>early</sub> resulted in a higher total number of TCs in the paracortex. B) Effector TCs that exited the paracortex were influenced more by SP<sub>early</sub> than by V<sub>max</sub>, negatively correlating with SP<sub>early</sub> (R² = 0.92, p<0.005). SP<sub>early</sub> negatively correlated with both (C) CD8<sup>+</sup> TCs exited (R² = 0.92, p<0.005) and (D) CD4<sup>+</sup> TCs exited (R² = 0.91, p<0.005). E) The mean number of cognate TCs contacted increased as SP<sub>early</sub> was lowered to 0.1 at each value of V<sub>max</sub> but overall decreased with V<sub>max</sub>. F) At a smaller maximal swelling, a higher TC to grid ratio was maintained but this did not translate to more or less effector TC production. G) Reduced SP<sub>early</sub> resulted in higher overall rates of TC egress in week 2 of the simulation.
Figure 12. Varying LN shutdown by modulating inflammation-induced S1P₁r downregulation (SP_{inflam}). Modulation of S1P₁r down-regulation in response to initial detection of inflammatory signals from no downregulation (SP_{inflam}=1) to 90% downregulation (SP_{inflam}=0.1), A) total TC number decreases several fold but (B) TC activation increases ($R^2 = 0.83$, $p=0.01$) while (C) Effector TC production shows no trend. D) Non-cognate TCs contact more DCs as SP_{inflam} is increased ($R^2=0.93$, $p=1.7\times10^{-3}$).