

# Notch signaling and taxis mechanisms regulate early stage angiogenesis: A mathematical and computational model

Short title: Notch signaling and taxis mechanisms regulation of early stage angiogenesis

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## Abstract

During angiogenesis, new blood vessels sprout and grow from existing ones. This process plays a crucial role in organ development and repair, in wound healing and in numerous pathological processes such as cancer progression or diabetes. We present here a mathematical model of early stage angiogenesis that permits to explore the relative importance of mechanical, chemical and cellular cues. Endothelial cells proliferate and move over an extracellular matrix by following external gradients of Vessel Endothelial Growth Factor, adhesion and stiffness, which are incorporated to a Cellular Potts model with a finite element description of elasticity. The dynamics of Notch signaling involving Delta-4 and Jagged-1 ligands determines tip cell selection and vessel branching. Through their production rates, competing Jagged-Notch and Delta-Notch dynamics determine the influence of lateral inhibition and lateral induction on the selection of cellular phenotypes, branching of blood vessels, anastomosis (fusion of blood vessels) and angiogenesis velocity. Anastomosis may be favored or impeded depending on the mechanical configuration of strain vectors in the ECM near tip cells. Numerical simulations demonstrate that increasing Jagged production results in pathological vasculatures with thinner and more abundant vessels, which can be compensated by augmenting the production of Delta ligands.

## Author Summary

Angiogenesis is the process by which new blood vessels grow from existing ones. This process plays a crucial role in organ development, in wound healing and in numerous pathological processes such as cancer growth or in diabetes. Angiogenesis is a complex, multi-step and well regulated process where biochemistry and physics are intertwined: with signaling in vessel cells being driven by both chemical and mechanical mechanisms that result in vascular cell movement, deformation and proliferation. Mathematical models have the ability to bring together these mechanisms in order to explore their relative relevance in vessel growth. In this work, we present a mathematical model of early stage angiogenesis that is able to explore the role of biochemical signaling and tissue mechanics. We use this model to unravel the regulating role of Jagged, Notch and Delta dynamics in vascular cells. These membrane proteins have an important part in determining the leading cell in each neo-vascular sprout. Numerical simulations demonstrate that increasing Jagged production results in pathological vasculatures with thinner and more abundant vessels, which can be compensated by augmenting the production of Delta ligands.

## Introduction

Angiogenesis is a process by which new blood vessels sprout and grow from existing ones. This ubiquitous phenomenon in health and disease of higher organisms [1], plays a crucial role in the natural processes of organ growth and repair [2],

wound healing [3], or inflammation [4]. Angiogenesis imbalance contributes to numerous malignant, inflammatory, ischaemic, infectious, and immune diseases [2, 5], such as cancer [6–9], rheumatoid arthritis [10], neovascular age-related macular degeneration [11], endometriosis [12, 13], and diabetes [14].

Either when a tissue is in hypoxia or during (chronic or non-chronic) inflammation, cells are able to activate signaling pathways that lead to the secretion of pro-angiogenic proteins. The Vascular Endothelial Growth Factor (VEGF) is one of these proteins and it is necessary and sufficient to trigger angiogenesis. Present in different isoforms, VEGF diffuses in the tissue, and is able to bind to extracellular matrix (ECM) components (its binding affinity is different for distinct VEGF isoforms), forming a well defined spatial concentration gradient in the direction of increasing hypoxia [15, 16]. When the VEGF molecules reach an existing vessel, they promote the dwindling of the adhesion between vessel cells and the growth of newer vessel sprouts. VEGF also activates the tip cell phenotype in the vessel endothelial cells (ECs) [17]. The tip cells grow filopodia rich in VEGF receptors, pull the other ECs, open a pathway in the ECM, lead the new sprouts, and migrate in the direction of increasing VEGF concentration [18]. Branching of new sprouts occur as a result of crosstalk between neighboring ECs [19].

As the new sprouts grow, ECs have to alter their shape to form a lumen connected to the initial vessel that is capable of carrying blood [20–24]. Moreover, in order for the blood to be able to circulate inside the new vessels, the growing sprouts have to merge either with each other or with existing functional mature vessels [25]. The process by which sprouts meet and merge is called anastomosis [25–29].

Nascent sprouts are then covered by pericytes and smooth muscle cells, which provide strength and allow vessel perfusion. Poorly perfused vessels may become thinner and their ECs, in a process that inverts angiogenesis, may retract to neighboring vessels leading to more robust blood circulation [30, 31]. Thus, the vascular plexus remodels into a highly organized and hierarchical vascular network of larger vessels ramifying into smaller ones [32].

Angiogenesis is therefore a multi-step, complex and well regulated process where biochemistry and physics are intertwined; with signaling in ECs being driven by both chemical and mechanical mechanisms that result in EC proliferation, mechanical deformation and cell movement.

In particular, the dynamical and biochemical processes that take place at the tip of every growing sprout are determinant for the growth, morphology and function of the resulting neo-vasculature. When an EC has the tip cell phenotype (which is triggered by the binding of VEGF to VEGF Receptor 2, VEGFR2) its membrane becomes rich in Delta-4 transmembrane proteins [19, 33]. These proteins bind the Notch transmembrane proteins in the neighboring cells triggering the Notch signaling pathway. The activation of this pathway down-regulates VEGFR2 and Delta-4, forcing the neighboring cells not to be in the tip cell phenotype, and to acquire the stalk cell phenotype [34]. Stalk ECs are characterised by a higher proliferation rate [18] triggered by both VEGF and by the tension exerted on them by the tip cell [35]. The sprouts are able to grow due to proliferation of the stalk ECs.

The ECs can interchange dynamically their phenotypes from tip to stalk. In fact, in the growing sprout the stalk ECs behind the tip cell are often able to overtake the tip cell and to take its place, thereby becoming tip cells and driving sprout elongation [36, 37]. This dynamic behavior ensures that there is always a cell at the front of the sprout with the tip phenotype capable of exerting a contractile force on the matrix, degrading and remodelling matrix fibres and opening a pathway for the sprout to grow.

EC metabolism is strongly connected with this cycling dynamics at the tip of sprouts [38], and it is determinant to vascular patterning, pruning and sprouting [39–41]. The ability of ECs to rearrange themselves is essential for vessel remodelling [30]. Moreover, this dynamics at the tip is only possible due to the regulation of VE-cadherin expression in ECs by the Notch signaling pathway [36, 42, 43]. When the Notch-driven tip-stalk pattern is absent (due to very high VEGF levels, for example) the EC rearrangement dynamics stops [43]. In that case the vessels become thicker and sprouting is severely hampered. Hence, the Notch signaling pathway is pivotal in determining the morphology of blood vessel networks.

Importantly, the dynamics of the ECs' phenotypes in a growing sprout can be rather complex. While at moderate values of VEGF lateral inhibition by tip ECs can be observed [44], at higher VEGF concentrations the situation is different. Recently it has been experimentally observed that high levels of VEGF lead to synchronisation of phenotypes between cells at the sprout [45]. This phenomenon had first been suggested by theoretical models [17]. The model suggested that ECs in a sprout under high VEGF levels initiate acquiring the tip cell phenotype simultaneously, and then all simultaneously trigger the lateral inhibition by the Delta-Notch signaling, losing the tip phenotype and moving towards the stalk phenotype, only for the process to start again. Synchronised oscillatory behaviour in Delta-4 levels in EC cells has been observed under these conditions [45]. In this way, high VEGF hinders the

symmetry breaking needed for the lateral inhibition to take place in the sprout. 67

Recent mathematical models of Notch signaling in angiogenesis have also predicted states where the cells in the 68  
sprout are in a third intermediate state and neither in the tip nor in the stalk phenotype [46, 47]. The Jagged-1 69  
transmembrane protein is an important partner in the regulation of the Notch signaling in angiogenesis, and its 70  
introduction in the computational models permit to predict these intermediate EC phenotypes [47–50]. 71

Jagged-1 is a ligand of Notch and competes with Delta-4 in angiogenesis [51]. Experiments have shown that 72  
when the lateral inhibition pattern induced by Delta-Notch signaling is present, the levels of Jagged-1 follow the EC 73  
phenotype: they are lower in tip cells and higher in stalk cells (contrary to what happens with the levels of Delta- 74  
4) [52]. However, ECs are able to control independently the levels of Jagged-1 (for example by reaction with proteins 75  
of the Fringe family [51]), and therefore they are able to control the sensitivity to Notch-mediated lateral inhibition. 76  
Moreover, Jagged-1 also plays an important role in making the Notch mechanism capable of lateral induction, whereby 77  
a stalk EC may induce its neighbors to acquire a phenotype equal to its own [53]. For these motives, it is extremely 78  
important to understand the implications of Jagged-1 levels in sprouting angiogenesis. Mathematical models should 79  
integrate the knowledge of Delta-Notch-Jagged signaling with the dynamics of EC organization in a sprout to better 80  
understand how the communication between ECs in angiogenesis is mediated by Jagged-1. 81

Numerous mathematical models of angiogenesis study the growth of blood vessels and irrigation using continuum 82  
methods, cellular automata, and hybrid methods [35, 54–73]. Cellular Potts Models (CPM) [74, 75] of angiogenesis 83  
and vasculogenesis have been particularly successful in capturing vascular cell shape [76], vascular structure [65, 77] 84  
and in integrating the role of extra-cellular matrix (ECM) mechanics and structure [66, 78–80] in the development of 85  
the vasculature. 86

Many of these models use simplified models of the Notch pathway to determine the separation between sprouts [25, 87  
67, 69]. However, very detailed models of the Notch signaling pathway that integrate the dynamics of filopodia growth 88  
and of anastomosis have been developed [17, 81]. These detailed models also shed light into the regulation of VE- 89  
Cadherin by Delta-Notch [36, 43] and into the coupling between EC metabolism, Delta-Notch and EC rearrangement 90  
dynamics at the tip [38]. 91

Moreover, cell based mathematical models that include Jagged-1 and Fringe have been developed in the contexts 92  
of cell differentiation [48–50] and angiogenesis [47]. However, these models of sprouting angiogenesis use a fixed 93  
geometry of a linear array of cells, without taking into account that ECs in a sprout are elongated and have many 94  
neighbors, and that they move and proliferate. Therefore, to describe the regulating effect of Jagged-1 in the 95  
sprouting dynamics we need to integrate dynamical models that take into account Jagged-1 with a CPM that takes 96  
into account cell shape and movement. In the present paper, we carry out this integration process for angiogenesis in 97  
the early stage, before sprouts form a lumen, become perfused and can regress. We use a CPM that incorporates cell 98  
motion following increasing gradients of VEGF (chemotaxis), of adhesion to substrate (haptotaxis) and of substrate 99  
stiffness (durotaxis), as well as cell proliferation and the Notch signaling pathway. This model will permit to explore 100  
the relative importance of mechanical, chemical and cellular cues in angiogenesis. 101

The section Mathematical Model describes the CPM coupled with the Delta-Notch-Jagged dynamics. In the 102  
section Results and Discussion, we present the results of the simulation and how Jagged-1 determines sprouting 103  
dynamics. Finally, in the last section we draw the conclusions of the manuscript. 104

## Mathematical model 105

The mathematical model consists of a CPM in which the dynamics of the Notch signaling pathway in endothelial 106  
cells selects tip and stalk ECs. Tip ECs move by chemotaxis, haptotaxis and durotaxis and stalk cells proliferate. 107  
Vessel branching and anastomosis appear as a result of combined cell signaling, mechanical and chemical taxis. 108

### Cellular Potts model 109

**Square grid.** We consider a square domain  $\Omega$  of side  $L$  with grid points  $(x_i, y_j)$ , where  $x_i = ih$ ,  $y_j = jh$  with 110  
 $i, j = 0, \dots, M - 1$ ,  $h = L/(M - 1)$ , and  $M$  is the number of nodes on a side of the square. The square contains 111  
 $M \times M$  grid points and  $(M - 1)^2$  elementary squares (pixels), each having an area  $L^2/(M - 1)^2$ . To enumerate 112  
nodes, we use left-to-right, bottom-to-top order, starting from node 0 on the bottom left corner of the square and 113  
ending at node  $M^2 - 1$  on the rightmost upper corner. In numerical simulations, we use  $L = 0.495$  mm. 114

**Objects, spins and Metropolis algorithm.** Pixels  $\mathbf{x}$  can belong to different objects  $\Sigma_\sigma$ , namely ECs, and ECM. The field (called spin in a Potts model)  $\sigma(\mathbf{x})$  denotes the label of the object occupying pixel  $\mathbf{x}$  [74]. Each given spin configuration for all the pixels in the domain has an associated energy  $H(\{\sigma(\mathbf{x})\})$  to be specified below. At each Monte Carlo time step (MCTS)  $t$ , we select randomly a pixel  $\mathbf{x}$ , belonging to object  $\Sigma_\sigma$ , and propose to copy its spin  $\sigma(\mathbf{x})$  to a neighboring (target) pixel  $\mathbf{x}'$  that does not belong to  $\Sigma_{\sigma(\mathbf{x})}$ . The proposed change in the spin configuration (spin flip) changes the configuration energy by an amount  $\Delta H|_{\sigma(\mathbf{x}) \rightarrow \sigma(\mathbf{x}' )}$ , and it is accepted with probability (Metropolis algorithm) [66, 74]

$$P(\sigma(\mathbf{x}) \rightarrow \sigma(\mathbf{x}'))(t) = \begin{cases} e^{-\Delta H|_{\sigma(\mathbf{x}) \rightarrow \sigma(\mathbf{x}')}/T}, & \Delta H|_{\sigma(\mathbf{x}) \rightarrow \sigma(\mathbf{x}')} > 0; \\ 1, & \Delta H|_{\sigma(\mathbf{x}) \rightarrow \sigma(\mathbf{x}')} \leq 0. \end{cases} \quad (1)$$

The temperature  $T > 0$  is measured in units of energy and it is related to an overall system motility. We have selected  $T = 4$  in our simulations.

**Energy functional.** The energy functional  $H$  is

$$H = \sum_{\Sigma_\sigma} \rho_{\text{area}} \left( \frac{a_\sigma - A_\sigma}{A_\sigma} \right)^2 + \sum_{\Sigma_\sigma} \rho_{\text{perimeter}} \left( \frac{p_\sigma - P_\sigma}{P_\sigma} \right)^2 + \sum_{\mathbf{x}, \mathbf{x}' \in \Omega'_x} \rho_{\text{adh}}^{\tau, \tau'} (1 - \delta_{\sigma, \sigma'}) + H_{\text{durot}} + H_{\text{chem}}. \quad (2)$$

Here the two first terms are sums over cells and the third one sums over all pixels. We have

- $a_\sigma$  is the area of the cell  $\Sigma_\sigma$ ,  $A_\sigma$  is the target area and  $\rho_{\text{area}}$  is the Potts parameter which regulates the fluctuations allowed around the target area. There are two cell types: non-proliferating tip and stalk cells with  $A_\sigma = 78.50 \mu\text{m}^2$  and proliferating cells with double target area,  $A_\sigma = 157 \mu\text{m}^2$ . The target radius of a proliferating cell is  $\sqrt{2}$  times that of a non-proliferating cell.
- $p_\sigma$  is the perimeter of the cell  $\Sigma_\sigma$ ,  $P_\sigma$  is the target perimeter and  $\rho_{\text{perimeter}}$  is the Potts parameter which regulates the fluctuations allowed around the target perimeter. The target perimeters are  $P_\sigma = 31.4 \mu\text{m}$  for non-proliferating cells, and thrice this,  $P_\sigma = 94.2 \mu\text{m}$ , for proliferating cells.
- The Potts parameter  $\rho_{\text{adh}}^{\tau, \tau'} \geq 0$  is the contact cost between two neighboring pixels. The value of this cost depends on the type of the object to which the pixels belong (cell or medium). Since  $\delta_{\sigma, \sigma'}$  is the Kronecker delta, pixels belonging to the same cell do not contribute a term to the adhesion energy.
- The net variation of the durotaxis term  $H_{\text{durot}}$  is [66]

$$\Delta H_{\text{durot}} = -\rho_{\text{durot}} g(\mathbf{x}, \mathbf{x}') \left( h(E(\epsilon_1)) \left( \mathbf{v}_1 \cdot \frac{\mathbf{x}' - \mathbf{x}}{|\mathbf{x}' - \mathbf{x}|} \right)^2 + h(E(\epsilon_2)) \left( \mathbf{v}_2 \cdot \frac{\mathbf{x}' - \mathbf{x}}{|\mathbf{x}' - \mathbf{x}|} \right)^2 \right), \quad (3)$$

where  $\rho_{\text{durot}}$  is a Potts parameter,  $g(\mathbf{x}, \mathbf{x}') = 1$  for extensions and  $g(\mathbf{x}, \mathbf{x}') = -1$  for retractions,  $\epsilon_1$  and  $\epsilon_2$  and  $\mathbf{v}_1$  and  $\mathbf{v}_2$  ( $|\mathbf{v}_1| = |\mathbf{v}_2| = 1$ ) are the eigenvalues and eigenvectors of the strain tensor  $\epsilon_T$ , respectively. They represent the principal strains and the strain orientation.  $\epsilon_T$  is the strain in the target pixel for extensions, and the strain in the source pixel for retractions.  $h(E) = 1 / (1 + \exp(-\omega(E - E_\theta)))$  is a sigmoid function with threshold stiffness  $E_\theta$  and steepness  $\omega$ .  $E(\epsilon) = E_0 (1 + (\epsilon/\epsilon_{\text{st}}) 1_{\epsilon \geq 0})$  is a function of the principal strains, in which  $E_0$  sets a base stiffness for the substrate,  $\epsilon_{\text{st}}$  is a stiffening parameter and  $1_{\epsilon \geq 0} = \{1, \epsilon \geq 0; 0, \epsilon < 0\}$ : strain stiffening of the substrate only occurs for substrate extension ( $\epsilon \geq 0$ ), whereas compression ( $\epsilon < 0$ ) does not stiffen or soften the substrate. We have used the parameter values:  $E_\theta = 15 \text{ kPa}$ ,  $E_0 = 10 \text{ kPa}$ ,  $\omega = 0.5 \text{ kPa}^{-1}$ , and  $\epsilon_{\text{st}} = 0.1$  [66].

- The variation of the chemotaxis term  $H_{\text{chem}}$  is

$$\Delta H_{\text{chem}} = -\rho_{\text{chem}} \frac{1}{1 + \alpha_{\text{chem}} C(\mathbf{x})} [C(\mathbf{x}') - C(\mathbf{x})], \quad (4)$$

where  $\rho_{\text{chem}} \geq 0$  is a Potts parameter,  $\alpha_{\text{chem}} = 0.3$  and  $C$  is the VEGF concentration in the corresponding pixel.

The values of the Potts parameters are listed in Table 1. They are chosen according to those proposed by Bauer *et al.* [76] and Van Oers *et al.* [66] and adjusted so as to make that every term of the net variation of the hamiltonian have the same order. The perimeter contribution, absent in Refs. [66, 76], is small compared to the other terms in Eq. (2), so that it only affects the computations in extreme cases (e.g., extremely thin cells, thin cells that stick to the blood vessel). We have added a factor in the chemotaxis term to regulate the fluctuation around the resting VEGF concentration. Note that if  $\alpha_{\text{chem}}$  is equal to zero, we recuperate the original term of Bauer *et al.* [76]. The proposed value,  $\alpha_{\text{chem}} = 0.3$ , is small.

Parameter	$\rho_{\text{area}}$	$\rho_{\text{perimeter}}$	$\rho_{\text{durot}}$	$\rho_{\text{chem}}$	$\rho_{\text{adh}}^{\tau, \tau'}$ (cell-cell)	$\rho_{\text{adh}}^{\tau, \tau'}$ (cell-ECM)
Value	9000	250	25	60000	8.25	16.50

**Table 1.** Dimensionless Potts parameters.

## Continuum fields at the extracellular scale

**VEGF concentration.** The VEGF concentration obeys the following initial-boundary value problem [76]:

$$\frac{\partial C(x, y, t)}{\partial t} = D_f \left( \frac{\partial^2 C(x, y, t)}{\partial x^2} + \frac{\partial^2 C(x, y, t)}{\partial y^2} \right) - \nu C(x, y, t) - G(x, y, C), \quad (x, y) \in \Omega, \quad t > 0, \quad (5)$$

$$C(0, y, t) = 0, \quad C(L, y, t) = S, \quad C(x, 0, t) = \frac{S}{L}x = C(x, L, t), \quad (x, y) \in \partial\Omega, \quad t > 0, \quad (6)$$

$$C(x, y, 0) = 0, \quad (x, y) \in \Omega. \quad (7)$$

In Eq. (5), the amount of VEGF bound by an EC per unit time is

$$G(x, y, C) = \begin{cases} \Gamma, & \text{if } \Gamma \leq \nu C(x, y) \text{ and } (x, y) \in \Sigma_{\text{EC}}, \\ \nu C, & \text{if } 0 \leq \nu C(x, y) < \Gamma \text{ and } (x, y) \in \Sigma_{\text{EC}}, \\ 0, & \text{if } (x, y) \notin \Sigma_{\text{EC}}, \end{cases} \quad (8)$$

where  $\nu = 1 \text{ h}^{-1}$  and  $\Gamma = 0.02 \text{ pg}/(\mu\text{m}^2 \text{ h})$  is the maximum amount of VEGF that it could be consumed by a cell per hour [62, 76]. Other values we use are  $D_f = 0.036 \text{ mm}^2/\text{h}$ ,  $\nu = 0.6498/\text{h}$ ,  $S = 0.0153 \text{ pg}/\mu\text{m}^2$  [76].

**Strains.** Following Ref. [66], we calculate the ECM strains by using the finite element method to solve the stationary Navier equations of linear elasticity:

$$\begin{aligned} K u &= f, & \text{in } \Omega, \\ u &= 0, & \text{in } \partial\Omega. \end{aligned} \quad (9)$$

Here  $K$  is the stiffness matrix,  $u$  is the array of the  $x$  and  $y$  displacements of all nodes and  $f$  is the array of the traction forces per unit substrate thickness exerted by the cells. For nodes outside ECs,  $f = 0$ . For nodes inside ECs, each component  $f_k = \mu_{\text{force}} \sum_j d_{kj} \delta_{\sigma_k, \sigma_j}$  represents the traction stress on the  $k$ th node,  $\mu_{\text{force}}$ , times the sum of the distances,  $d_{kj}$ , between the  $k$ th node and any node  $j$  in the same cell ( $\sigma_k$  is the label of the cell at which node  $k$  belongs).

The global stiffness matrix  $K$  is assembled from the stiffness matrices  $K_e$  of each pixel,

$$K_e = \int_{\Omega_e} B^T D B \, d\Omega_e, \quad (10)$$

in which

$$D = \frac{E}{1 - \nu^2} \begin{pmatrix} 1 & \nu & 0 \\ \nu & 1 & 0 \\ 0 & 0 & \frac{1}{2}(1 - \nu) \end{pmatrix}, \quad (11)$$

and  $B$  is the strain-displacement matrix for a four-noded quadrilateral pixel (finite element) [66].  $B$  is a  $3 \times 8$  matrix that relates the 8-component node displacement  $u_e$  of each pixel to local strains  $\epsilon$ ,

$$\epsilon = B u_e, \quad (12)$$

where  $\epsilon = (\epsilon_{11}, \epsilon_{22}, \epsilon_{12})$  is the 3-component column notation of the strain tensor

$$\epsilon_T = \begin{pmatrix} \epsilon_{11} & \epsilon_{12} \\ \epsilon_{12} & \epsilon_{22} \end{pmatrix}. \quad (13)$$

We have used the numerical values  $E = 10$  kPa,  $\nu = 0.45$ , and  $\mu_{\text{force}} = 1$  N/m<sup>2</sup>. With these definitions and the durotaxis term given by Eq. (3), ECs generate mechanical strains in the substrate, perceive a stiffening of the substrate along the strain orientation, and extend preferentially on stiffer substrate. The simulated ECs spread out on stiff matrices, contract on soft matrices, and become elongated on matrices of intermediate stiffness [66].

## Signaling processes

The Notch signaling pathway is activated when Notch (transmembrane receptor) belonging to a particular cell interacts with Delta-4 or Jagged-1 (transmembrane ligands) belonging to its neighboring cell (trans-activation), thereby releasing the Notch intracellular domain (NICD). NICD then enters the nucleus and modulates the expression of many target genes of the Notch pathway, including both the ligands Delta and Jagged. However, when Notch of a cell interacts with Delta or Jagged belonging to the same cell, no NICD is produced; rather, both the receptor (Notch) and ligand (Delta or Jagged) are degraded (cis-inhibition) and therefore the signaling is not activated. For a given cell  $i$  surrounded by other cells, the equations describing this pathway are [47]

$$\frac{dN_i}{dt} = r_N H^S(I_i, \lambda_{I,N}) - \{[k_C D_i + k_T D_{\text{ext}}(i)] H^S(I_i, \lambda_{D,F}) + [k_C J_i + k_T J_{\text{ext}}(i)] H^S(I_i, \lambda_{J,F}) + \gamma\} N_i, \quad (14)$$

$$\frac{dD_i}{dt} = r_D H^S(I_i, \lambda_{I,D}) H^S(V_i, \lambda_{V,D}) - [k_C N_i H^S(I_i, \lambda_{D,F}) + k_T N_{\text{ext}}(i) + \gamma] D_i, \quad (15)$$

$$\frac{dJ_i}{dt} = r_J H^S(I_i, \lambda_{I,J}) - [k_C N_i H^S(I_i, \lambda_{J,F}) + k_T N_{\text{ext}}(i) + \gamma] J_i, \quad (16)$$

$$\frac{dI_i}{dt} = k_T N_i [H^S(I_i, \lambda_{D,F}) D_{\text{ext}}(i) + H^S(I_i, \lambda_{J,F}) J_{\text{ext}}(i)] - \gamma_S I_i, \quad (17)$$

$$\frac{dV_{Ri}}{dt} = r_{V_R} H^S(I_i, \lambda_{I,V_R}) - k_T V_{Ri} V_{\text{ext}}(i) - \gamma V_{Ri}, \quad (18)$$

$$\frac{dV_i}{dt} = k_T V_{Ri} V_{\text{ext}}(i) - \gamma_S V_i. \quad (19)$$

Here,  $N_i$ ,  $D_i$ , and  $J_i$  are the number of Notch, Delta-4, and Jagged-1 proteins in the  $i$ th cell, respectively, at time  $t$ .  $I_i$ ,  $V_{Ri}$  and  $V_i$  are the number of NICD, VEGF receptor and VEGF molecules, respectively, that are in the  $i$ th cell at time  $t$ .  $r_N$ ,  $r_D$ ,  $r_J$ , and  $r_{V_R}$ , are the production rates of  $N$ ,  $D$ ,  $J$ , and  $V_R$ , respectively. The cis-inhibition and trans-activation rates are  $k_C$  and  $k_T$ , respectively, whereas  $\gamma$  and  $\gamma_S$  are degradation rates for  $N$ ,  $D$ ,  $J$ ,  $V_R$  and for  $I$ ,  $V$ , respectively. These parameters, their representative values and units are listed in Table 2.

Parameter	$r_N$	$r_D, r_J, r_{V_R}$	$k_C$	$k_T$	$\gamma$	$\gamma_S$
Value	1200	1000	$5 \times 10^{-4}$	$2.5 \times 10^{-5}$	0.1	0.5
Unit	molec/h	molec/h	(h molec) <sup>-1</sup>	(h molec) <sup>-1</sup>	h <sup>-1</sup>	h <sup>-1</sup>

**Table 2.** Rates appearing in Eqs. (14)-(19).

Outside the  $i$ th cell, the number of  $X$  molecules is

$$X_{\text{ext}}(i) = \frac{1}{P_i} \sum_{j \in (i)} P_{i,j} X_j, \quad (20)$$

where  $X = N, D, J$ , and  $j \in \langle i \rangle$  are the cells  $j$  sharing boundary of length  $P_{i,j}$  with cell  $i$ . The perimeter of cell  $i$ ,  $P_i$ , minus  $\sum_{j \in \langle i \rangle} P_{i,j}$  is the length of its boundary that is not shared with any other cell. Note that  $X_{\text{ext}}(i)$  is simply the sum of all  $X_j$  if the lengths  $P_{i,j}$  are all equal and  $P_i = \sum_{j \in \langle i \rangle} P_{i,j}$  because the whole boundary of cell  $i$  is shared with other cells.  $V_{\text{ext}}(i)$  is the number of VEGF molecules outside the  $i$ th cell that interact with VEGF receptor cells to produce VEGF molecules inside the  $i$ th cell. The external VEGF cells come from the continuum field  $C(x, y, t)$ , which diffuses from  $x = L$ . Let  $\mathbf{x}_i$  be the pixel of the  $i$ th cell that is closer to the hypoxic region. The number of external VEGF molecules in that pixel is  $C(\mathbf{x}_i, t)$  multiplied by the conversion factor  $\chi_V = N_A L^2 / [(N - 1)^2 M_V]$ , where  $M_V$  is the molecular weight of the VEGF molecules and  $N_A$  is the Avogadro number. We have used  $\chi_V = 1$ , which is representative of VEGF molecules with a large molecular weight. In the numerical simulation,  $C$  is known in the grid points and its value at a pixel should be the average value of the four grid points of the pixel. Since these values are quite similar, we adopt the value of  $C$  at the bottom left grid point of the pixel  $\mathbf{x}_i$  as  $C(\mathbf{x}_i, t)$ .

Parameter	$\lambda_{I,N}, \lambda_{V,D}, \lambda_{I,J}$	$\lambda_{I,D}, \lambda_{I,V_R}$	$\lambda_{D,F}$	$\lambda_{J,F}$	$n_N, n_D, n_V, n_{V_R}$	$n_J$	$n_F$	$I_0, V_0$	$\chi_V$
Value	2.0	0.0	3.0	0.3	2.0	5.0	1.0	200	1.0

**Table 3.** Dimensionless parameters appearing in the Hill functions.  $I_0$  and  $V_0$  are activation numbers of NICD and VEGF molecules, respectively, and  $\chi_V$  is the conversion factor.

The shifted, excitatory and inhibitory Hill functions appearing in Eqs. (14)-(19) are:

$$H^S(\xi, \lambda_{\eta,\zeta}) = H^-(\xi) + \lambda_{\eta,\zeta} H^+(\xi), \quad (21)$$

$$H^-(\xi) = \frac{1}{1 + \left(\frac{\xi}{\xi_0}\right)^{n_\zeta}}, \quad H^+(\xi) = 1 - H^-(\xi), \quad (22)$$

where  $H^S$  is excitatory for  $\lambda_{\eta,\zeta} > 1$  and inhibitory for  $\lambda_{\eta,\zeta} \leq 1$ . In Eqs. (21)-(22),  $\xi = V, I$ ,  $\eta = I, V, D, J$ , and  $\zeta = N, D, J, V_R, F$  (the subscript  $F$  refers to Fringe, cf. [47]). The dimensionless parameters  $n_\zeta$  and  $\lambda_{\eta,\zeta}$  appearing in the Hill functions are listed in Table 3.

Variable	$N_i, D_i, J_i, N_{\text{ext}}, D_{\text{ext}}, J_{\text{ext}}$	$I_i$	$V_{Ri}$	$V_i$	$V_{\text{ext}}$	$t$
Scale	$\sqrt{r_D/k_C}$	$(k_T r_D)/(k_C \gamma_S)$	$r_{V_R}/\gamma$	$V_0$	$6V_0$	$1/\sqrt{k_C r_D}$
Value	$\sqrt{2} \times 10^3$	$10^2$	$10^4$	$2 \times 10^2$	$12 \times 10^2$	$\sqrt{2}$
Unit	molec	molec	molec	molec	molec	h

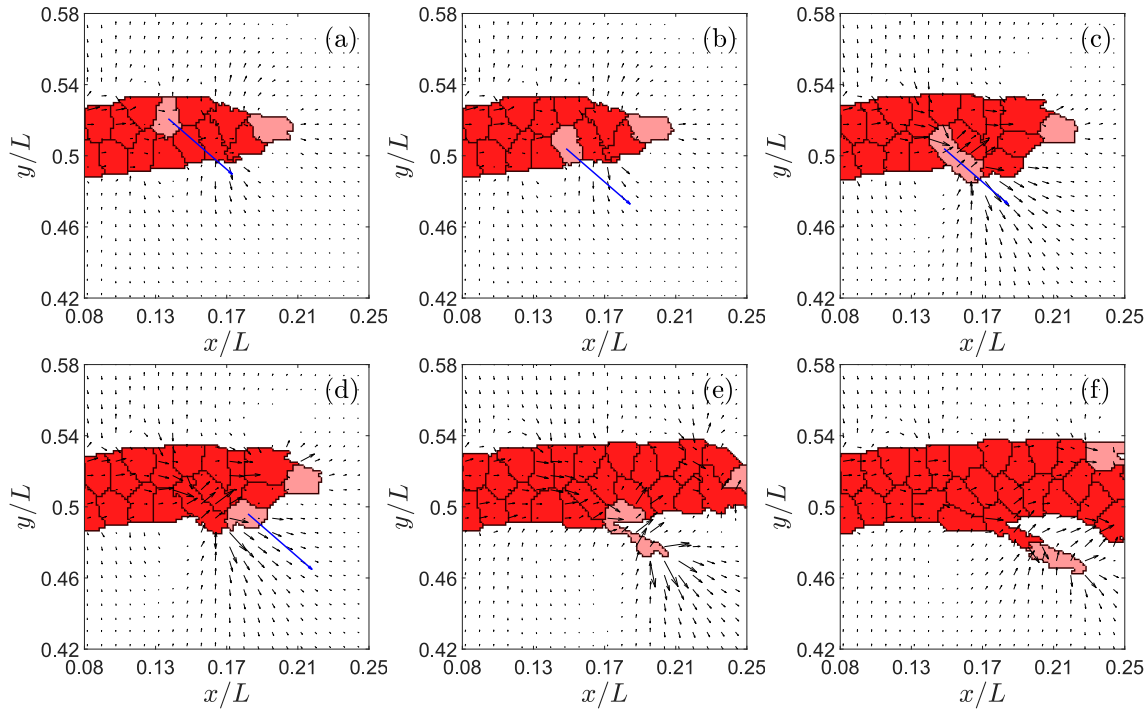
**Table 4.** Units for nondimensionalizing the Notch equations (14)-(19).

## Cell types, proliferation, branching and anastomosis

**Cell types.** In the model, ECs may be on a tip, hybrid or stalk cell phenotype. In nature, tip cells are characterized by having high levels of Delta-4, VEGFR2, and active VEGF signaling (i.e., high levels of VEGF internalization). They develop filopodia and migrate along the VEGF-A gradient, leading the formation of new branches. Delta-4 proteins at tip cell membranes inhibit the neighboring cells (due to lateral inhibition) to adopt a tip phenotype, thereby forcing them to become stalk cells (with low Delta-4, VEGFR2 and internalized VEGF).

Likewise, in our model, tip cells are distinguished by the number of VEGF molecules they possess. Therefore, a cell that has  $V$  larger than all its neighbors and  $V > 0.5 \max_i V_i(t)$  will acquire the tip cell phenotype and be very motile. To simulate this, tip cells are able to follow the mechanical and chemical cues on the environment, having  $\rho_{\text{chem}} \neq 0$  and  $\rho_{\text{durot}} \neq 0$ . On the other hand, stalk cells are not motile and they have  $\rho_{\text{chem}} = \rho_{\text{durot}} = 0$  in the model (except when they undergo proliferation, see below). Stalk cells, by virtue of the lateral induction, characteristic of Notch-Jagged signaling, are able to induce neighboring cells to adopt a stalk cell phenotype, by promoting a decrease of internal VEGF in them.

In our model we track the cells belonging to each growing vessel. A new sprouting vessel can be formed when a stalk cell acquires the tip phenotype. This cell can then become the leading cell of a new vessel that branches out from out the old one. This is illustrated in Fig. 1. If the levels of VEGF inside the tip cells that lead an active growing branch drop to values in the interval  $0.2 \max_i V_i(t) < V < 0.5 \max_i V_i(t)$ , these cells will be in the hybrid phenotype.



**Figure 1.** Example of tip cell exchange and branching in the direction of the blue arrow. Times in MCTS are: 422, (b) 423, (c) 460, (d) 461, (e) 545, (f) 630. The black arrows in this figure represent the directions of largest eigenstrain and, therefore, they point to the likeliest direction of EC motion. The blue arrows indicate the actual direction of motion of a selected tip cell (marked in pink color) for the simulation we have carried out.

In spite of the lower amount of Delta-4, VEGFR2 and VEGF, these cells remain with the tip cell characteristics and are able to lead the sprout. Similarly, stalk cells whose internal VEGF increases to the same range acquire the hybrid phenotype and can lead a sprout. The number of cells in the hybrid phenotype is only appreciable for larger Jagged production rates. 216  
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**Branching.** When a stalk cell (which does not border an existing tip cell) acquires the tip cell or the hybrid tip/stalk cell phenotype, this event will lead to the creation of a new active sprouting branch depending on its localization within the existing branch and on its moving direction. 220  
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To create a new branch, the boundary of the tip cell must touch the ECM. Moreover, let  $\mathcal{P}$  be the set of  $n_{\mathcal{P}}$  ECM pixels that have boundary with the branching tip cell. For each pixel  $\mathbf{x}_p \in \mathcal{P}$ , let the *strain vector* be  $\mathbf{v}_p = \epsilon_j \mathbf{v}_j$ , where  $\epsilon_j$  is the largest eigenstrain at pixel  $\mathbf{x}_p$  and  $\mathbf{v}_j$  is the corresponding unit eigenvector, as defined after Eq. (3). The average modulus and argument for the branching cell  $i$  are 223  
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$$\Lambda_i = \frac{1}{n_{\mathcal{P}}} \sum_{p \in \mathcal{P}} |\mathbf{v}_p|, \quad \theta_i = \frac{1}{n_{\mathcal{P}}} \sum_{p \in \mathcal{P}} \text{Arg } \mathbf{v}_p. \quad (23)$$

Let us also assume that the gradient of the chemotactic factor  $C$  forms an angle  $\Theta$  with the  $x$ -axis. The new tip cell will branch out, creating a new vessel, if the direction given by  $\theta_i$  points in the direction of the ECM and if  $-\pi/2 < \theta_i - \Theta < \pi/2$ . For other values of  $\theta_i$ , the tip cell does not leaves the parent vessel, since the chemotactic term of Eq. (4) opposes branching. In those cases, the direction given by  $\theta_i$  points to another cell, not to the ECM. To facilitate branching computationally, we directly exchange the new tip cell with this neighboring cell (see Figs. 1(a) and 1(b)). These exchanges may continue in successive MCTS until the new tip cell reaches a boundary of the blood vessel for which the direction given by  $\theta_i$  points to the ECM, as shown in Fig. 1. These cell exchanges in 2D mimic the climbing motion of the new tip cell over the parent vessel in a 3D geometry without merging with it. 227  
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We set the branching process to take at least 400 MCTS (incubation time). During this time we implement a persistent motion of the new tip cell in the direction marked by the angle  $\theta_i$ . During this incubation period, and to provide a good separation from the parent vessel we permit the tip cell to proliferate once (see Figs. 1(e) and 1(f), and see below). After this time, the dynamics of the branching vessel follows the same rules as that of any other actively sprouting vessel.

**Cell proliferation.** Endothelial cell proliferation in sprouting angiogenesis is regulated by both mechanical tension and VEGF concentration. In sprouting angiogenesis the tip cell creates tension in the cells that follow its lead. On those first stalk cells, this tension produces strain that triggers cell proliferation, if VEGF concentration is high enough [35]. Therefore, in our model, for each active sprouting vessel, one of the stalk cells that is in contact with a tip cell is randomly chosen to undergo proliferation. Only one cell per sprout proliferates, and it takes an average of 20 MCTS to do so. Tip cells in the model cannot proliferate, except only once when they start a new branch.

Once a stalk cell attached to a tip cell has been randomly selected as a proliferating cell, its target area in the CPM is set to become twice the size, whereas its target perimeter is set to a value three times that of non-proliferating cells. This cell will then grow in successive MCTS until it reaches this large target area. Then the cell proliferates if the following three conditions hold: (i)  $C(\mathbf{x}_i, t) > \psi_p$  (external VEGF surpasses a threshold), (ii) the cell belongs to an active blood vessel with cell proliferation, and (iii) the cell is not surrounded completely by other cells. Failure to meet one of these conditions precludes proliferation. If the three conditions are met, we use the unsupervised machine learning algorithm *K-means clustering* to split the cell. This algorithm calculates the Euclidean distance of each pixel in the cell to the centroid of two groups of pixels and corrects the centroids until the two pixel groups are balanced. These two groups comprise the new cells. Provided the daughter cells share boundary with the tip cell, one of them is randomly chosen to retain the ability to proliferate but the other cell does not proliferate. If the daughter cells do not share boundary with the tip cell, they both become non-proliferating and a different cell that shares boundary with the tip cell is randomly chosen to become a proliferating cell. Non-proliferating stalk cells have  $\rho_{\text{chem}} = \rho_{\text{durot}} = 0$ , whereas proliferating stalk cells are affected by chemo and durotaxis and their  $\rho_{\text{chem}}$  and  $\rho_{\text{durot}}$  values are as in Table 1. Setting  $\rho_{\text{chem}} = \rho_{\text{durot}} = 0$  for all proliferating stalk cells impedes branching of new sprouting vessels that are unable to leave the neighborhoods of their parent vessels.

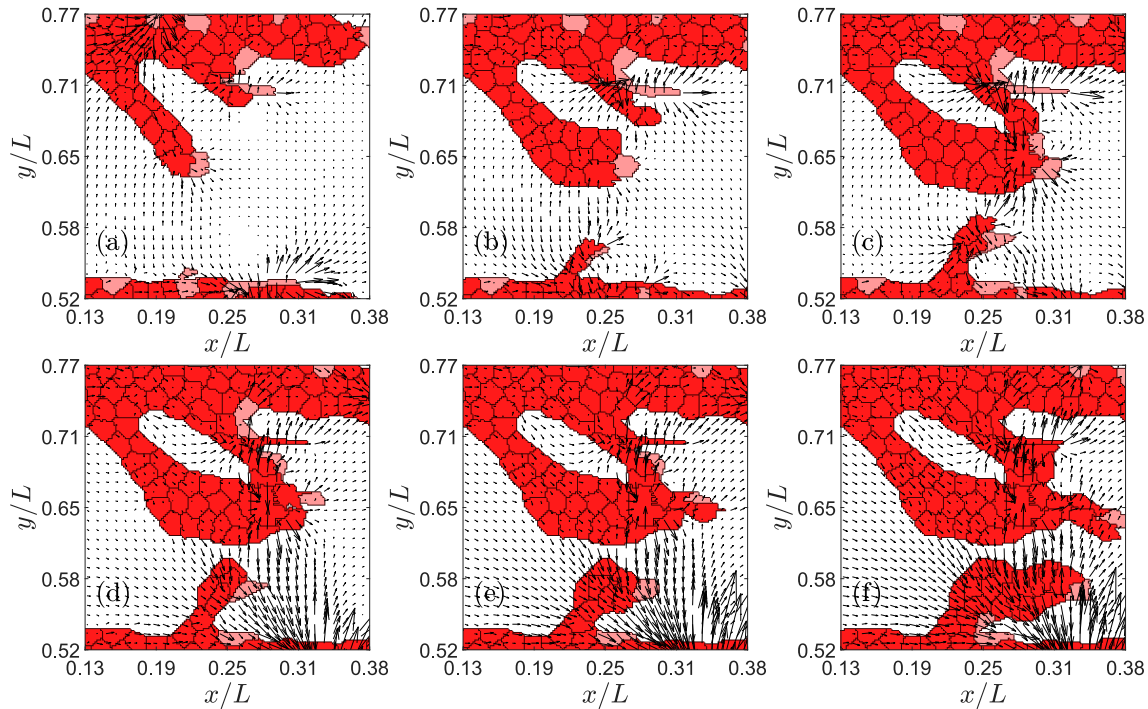
**Anastomosis.** When an active sprouting blood vessel merges with another active sprouting vessel, i.e. during anastomosis, one of them becomes inactive. If the collision occurs between tip cells of two different vessels, one vessel is randomly chosen to become inactive. If one tip cell merges with a stalk cell of a different active sprouting vessel, the vessel to which the tip cell belongs becomes inactive. The cells of an inactive vessel do not proliferate or branch, although they continue to undergo Notch signaling dynamics.

## Results and discussion

We have run our simulation model for a simple slab geometry and different conditions. A primary vessel is supposed to be along the  $y$  axis. The initial VEGF concentration  $C(\mathbf{x}, 0)$  is independent of  $y$  and decays linearly in  $x$  from  $x = L$  to  $x = 0$ . Thus, chemotaxis pushes tip cells towards the vertical line at  $x = L$ .

**Cellular mechanics and anastomosis.** We have found that cellular mechanics significantly controls anastomosis. The arrows in Fig. 2 are directed along the strain vector (eigenvector corresponding to the largest eigenstrain and having length equal to that eigenstrain). According to Eq. (3), the arrows indicate the likeliest direction in which ECs will move. The snapshots depicted in Fig. 2 show examples of successful and frustrated anastomosis and branching of advancing blood vessels. A tip cell leads successful branching from the blood vessel at the bottom of Fig. 2(a), as shown by Panels (b) and (c). Meanwhile, two blood vessels that sprout from the blood vessel at the top of Fig. 2(a) successfully anastomose as shown in Fig. 2(c). Notice that the strain vectors show the path of the approaching vessels until they anastomose. However, the branches arising from the two lowest vessels in Fig. 2(a) do not anastomose. They approach each other in Fig. 2(c) but the strain vectors pull them away from each other and anastomosis is frustrated, as shown in Figs. 2(d) and 2(e) [66, 82, 83].

Tip cells have higher levels of VEGF and their motion follows stiffness, chemical and adhesion gradients, as expected from the model. In successful anastomosis, one tip cell is directed by the strain vector to one actively



**Figure 2.** Example of successful and frustrated anastomosis. Times in MCTS are: (a) 751, (b) 851, (c) 951, (d) 1051, (e) 1101, (f) 1201. Tip cells are pink.

sprouting vessel. When it makes contact, it fuses with that vessel. After that, the VEGF in the tip cell decreases and it becomes a stalk cell. 282  
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**Jagged–Delta dynamics and sprouting.** Jagged and Delta dynamics determine sprouting [47, 48]. Studies of Notch signaling in one cell driven by external Jagged and Delta molecules show that the phenotype of a tip cell changes to hybrid tip/stalk and then to stalk cell as the external Delta concentration surpasses successive thresholds (cf. Fig. 3 of Ref. [47]). The thresholds depend on the Jagged production rate. Lateral induction works similarly for one cell driven by external Jagged molecules: tip cells change to hybrid tip/stalk and stalk phenotypes as the external Jagged concentration surpasses successive Delta-dependent thresholds [47, 48]. Simulations of our model illustrate the effects of J-N and D-N signaling combined with chemo-, hapto- and durotaxis. Figure 3 shows that increasing the Jagged production rate  $r_J$  yields smaller branching blood vessels, thereby decreasing the irrigation of the hypoxic region. Furthermore, sprouting is accelerated as the Jagged production augments: thinner and less efficient sprouts are formed faster as  $r_J$  increases. Stalk cells proliferate on advancing sprouts. Thus, increasing the number of tip cells leading sprouts results in increasing cell proliferation and a more rapid sprout advance. This behavior agrees with the sketch in Fig. 5A of Ref. [47], which indicates that pathological angiogenesis is obtained when there is an excess Jagged production. The sprouts in physiological angiogenesis are thicker and advance more slowly than the more abundant and thinner sprouts in pathological angiogenesis, as shown in Fig. 3. 284  
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The Delta production rate  $r_D$  acts opposite to  $r_J$ . High and intermediate levels of  $r_D$  ensure physiological angiogenesis, whereas the numbers of the hybrid tip/stalk cells increase for low levels of  $r_D$ . In more detail, we observe that, for  $r_J = 500$  molec/h and  $r_D = 1000$  molec/h, Figs. 4(a) and 4(b) show a gap between the VEGF of stalk and tip cells: the content of  $V$  is very low for stalk cells. It increases monotonically with  $V_{\text{ext}}$  and  $J$  for hybrid-tip and tip cells. Fig. 4(b) also shows that tip cells and hybrid tip-stalk cells have larger  $J$  than stalk cells. As  $r_J$  increases, at  $r_J = 2000$  molec/h, the hybrid-tip cells have proliferated and bridge the gap in  $V$ , as depicted in Figs. 4(c) and 4(d). Fig. 4(d) indicates that  $J$  is smaller for the tip cells at large Jagged production rates, which is consistent with lateral induction of stalk phenotype by stalk cells with large  $J$  values [47]. For large  $r_J$ , tip cells have less Jagged ( $J \approx 10$  and  $V > 2$ ) than other cell types ( $J$  between 10 and 15 and  $V < 2$ ), as shown in Fig. 4(d). 298  
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Fig. 4(e) and 4(f) show that the Delta production rate  $r_D$  acts in opposition to  $r_J$ . At  $r_D = 7500$  molec/h, Fig. 4(e), there is again a gap between the VEGF of stalk and tip cells. At this large Jagged production rate, tip cells have lower Jagged than stalk cells, as depicted in Fig. 4(f), which is similar to Fig. 4(d). However, Fig. 4(f) exhibits a gap between the maximum value of  $J$  for tip cells and the values of  $J$  for stalk cells, as compared to Fig. 4(d).

**Jagged-Delta dynamics and anastomosis.** What is the effect of modifying J-N and D-N signaling on angiogenesis? Figs. 3 and 5 show the effects of lateral inhibition by D-N signaling versus lateral induction by J-N signaling. Increasing the Jagged production rate produces more hybrid tip/stalk cells and more sprouts, as shown by Figs. 3 and, for the higher content of hybrid tip/stalk cells, by Figs. 4(c) and 4(d). However, for a high Jagged production rate, increasing the Delta production rate favors lateral inhibition by tip cells, which eventually decreases the number of new sprouts, makes anastomosis less frequent, as illustrated by Figs. 3(g), 3(h), 3(i), and 5.

Fig. 6 shows the concentrations of  $N$ ,  $V$ ,  $J$  and  $D$  for a developed angiogenic network for several values of the Jagged-1 and Delta-4 production rates. We observe that tip cells have large values of  $V$  and  $D$  for both normal ( $r_J = 500$  molec/h) and high ( $r_J = 2000$  molec/h) Jagged production rates, cf. Figs. 6(g)-(i) and 6(m)-(o). These figures highlight the role of lateral inhibition on stalk cells that are neighbors of tip cells. For large  $r_J$ , Figs. 6(k) and 6(l) show that stalk cells clearly have larger values of  $J$ , thereby illustrating the more important role of lateral induction. For large  $r_J$  and moderate  $r_D$ , Fig. 6(h) exhibits a larger number of cells with intermediate values of their internal VEGF, which shows the abundance of the hybrid tip/stalk cell phenotype. This is not the case for lower Jagged production rate as shown by the VEGF content in Figs. 6(g) and, for higher Delta production rate, in Fig. 6(i). As explained before and as shown by comparing Figs. 4(b) to 4(d) and 4(f), stalk cells have a smaller value of  $J$  than tip or hybrid tip/stalk cells at smaller Jagged production rates. In these cases, lateral inhibition by D-N signaling is more important. Figs. 6(m), 6(n) and 6(o) show that the  $D$  level of stalk cells is much reduced as compared with that of neighboring tip cells. Increasing the production rate of Delta-4 restores the morphology of the advancing normal vasculature to angiogenesis with high Jagged production rate, as shown by a comparison of Fig. 6(n) to Figs. 6(o) and 6(m).

Figure 7 further shows the effect of varying the production rates of Jagged and Delta on the advance and morphology of the vascular plexus. With respect to the simulations in Figs. 1 and 2 for standard values of  $r_J$  and  $r_D$ , increasing the production of Jagged, as shown in Fig. 3, produces more tip cells that run faster, cf. Figs. 7(a) and 7(b). Thus, lateral induction mediated by Jagged accelerates the advance of vasculature and increases the number of blood vessels by creating more hybrid tip/stalk cells, as explained before in relation to Figs. 3 and 4. If we keep constant  $r_J$  and increase the Delta production rate, lateral inhibition by tip cells becomes stronger, cf. Fig. 3. Then the number of tip cells decreases whereas the vasculature advances only slightly faster because angiogenesis and anastomosis diminish compared with the case of smaller  $r_D$ , cf. Fig. 7(c) and 7(d).

## Conclusion

The mathematical model of angiogenesis presented here illustrates the relative importance of mechanical, chemical and cellular cues when they are all considered simultaneously. Given a proliferation rate and a VEGF gradient on a homogeneous extracellular matrix, competing J-N and D-N dynamics determine the influence of lateral inhibition and lateral induction on tip cell selection, branching, anastomosis and speed of angiogenesis. Anastomosis is driven by chemotaxis. Cellular motion is informed by haptotaxis and durotaxis. However, anastomosis may be favored or impeded depending on the mechanical configuration of strain vectors in the ECM near tip cells. Notch signaling determines tip cell selection and vessel branching. Lateral induction by stalk cells and lateral inhibition by tip cells are informed by competing Jagged-Notch and Delta-Notch dynamics in manners that depend quantitatively on the Delta and Jagged production rates. In particular, the numerical simulations of our model predict the following effects of the production rates. Increasing the production rate of Jagged favors lateral induction of stalk cells, which yields more hybrid tip/stalk cells and a thinner vasculature that advances faster. On the other hand, increasing the production rate of Delta lowers the number of tip cells by lateral inhibition of stalk cells. Then there are less sprouts and anastomosis is less frequent while the advance of the vascular plexus is only slightly faster. Our numerical simulations illustrate the regulating role of Notch-Jagged-Delta signaling in the velocity and morphology of angiogenic vasculature. An imbalance of the Jagged production, so that there is more Jagged and increased lateral induction of stalk cells, results in anomalous thinner sprouts and faster angiogenesis. This may be corrected by

increasing the Delta-4 production rate, which boosts lateral inhibition of tip on stalk cells, diminishes the number of tips and slows down somewhat angiogenesis.

Our 2D model of early stage angiogenesis needs to be extended in several directions to be made more realistic and to account for later stages of angiogenesis. The extension of the model to three dimensional configurations is straightforward although it requires more computing power. While we have studied relatively short distances between the primary vessel and the target hypoxic region, we need to consider larger systems to be able to do statistical studies of vessel numbers and their width. To move toward later stages of the formation of an advancing vascular plexus, we need to add lumen formation [24] and blood circulation to the model [31]. These processes will allow us to tackle the concurrent sprouting and anastomosis on the front of the advancing vascular plexus and the pruning of poorly perfused sprouts on its back [31,32].

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## Supporting Information

S1 Text. Supporting movies: .

## Author Contributions

Conceived and designed the work: RV MC RDMT LLB. Implemented and performed the numerical simulations: RV MC. Analyzed the data: RV MC RDMT LLB. Wrote the paper: RV MC RDMT LLB.

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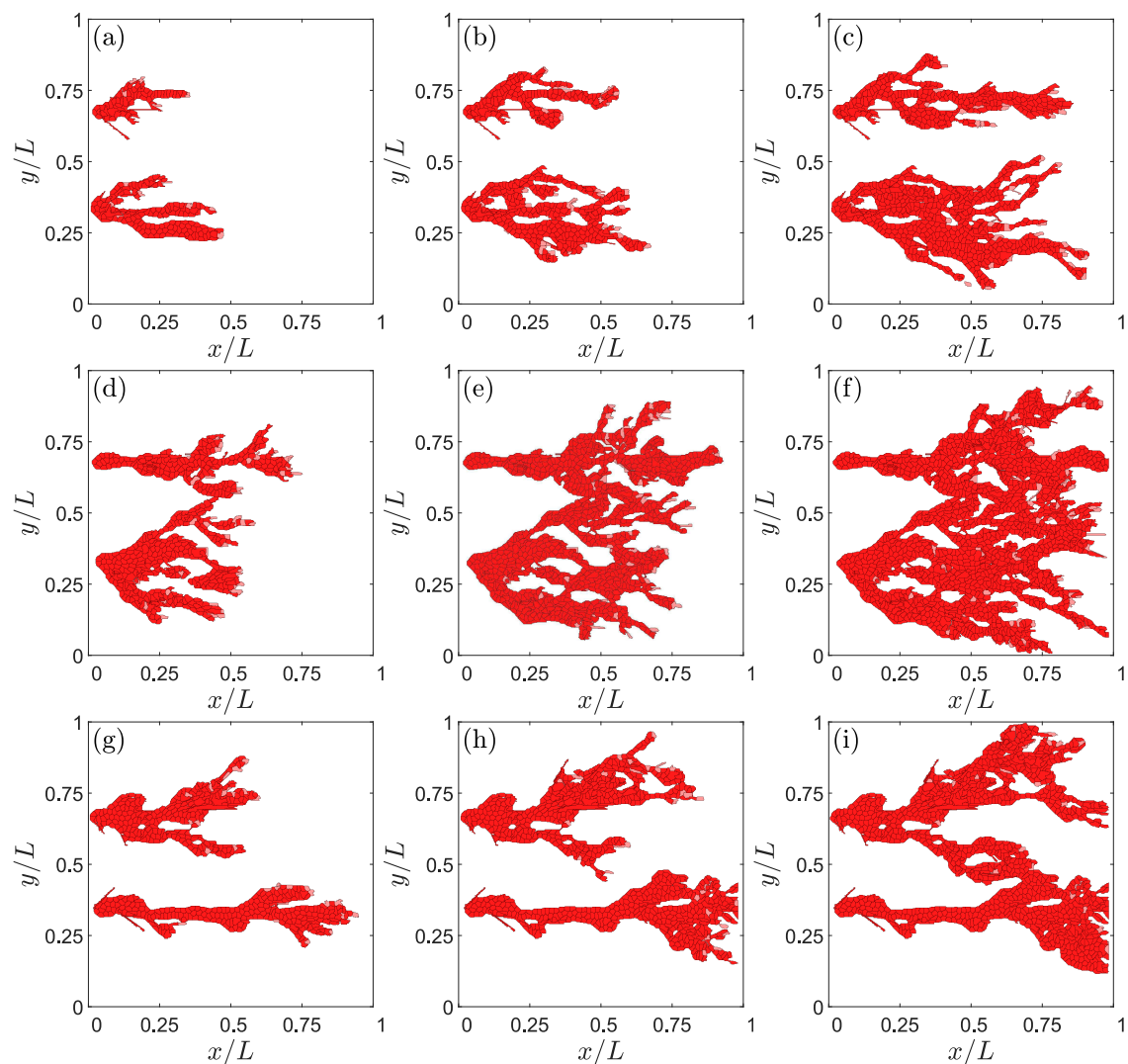
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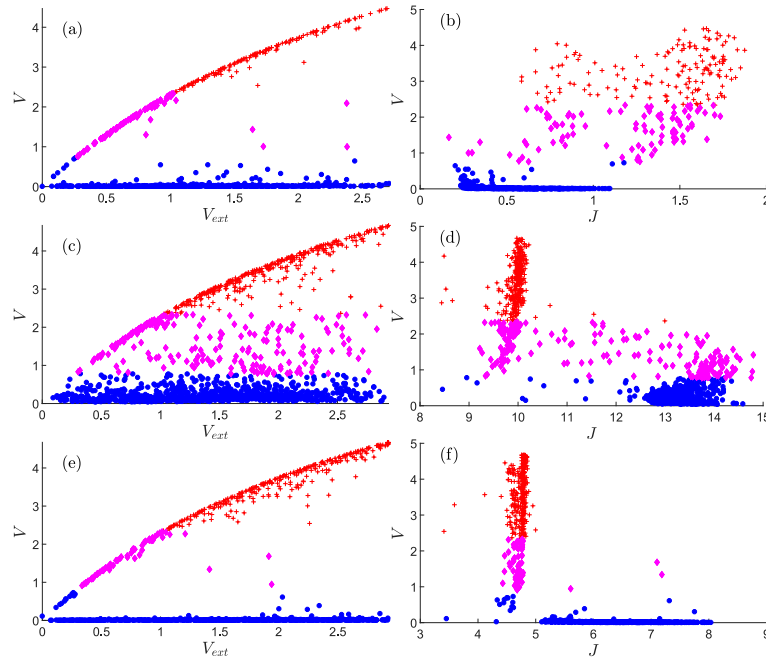
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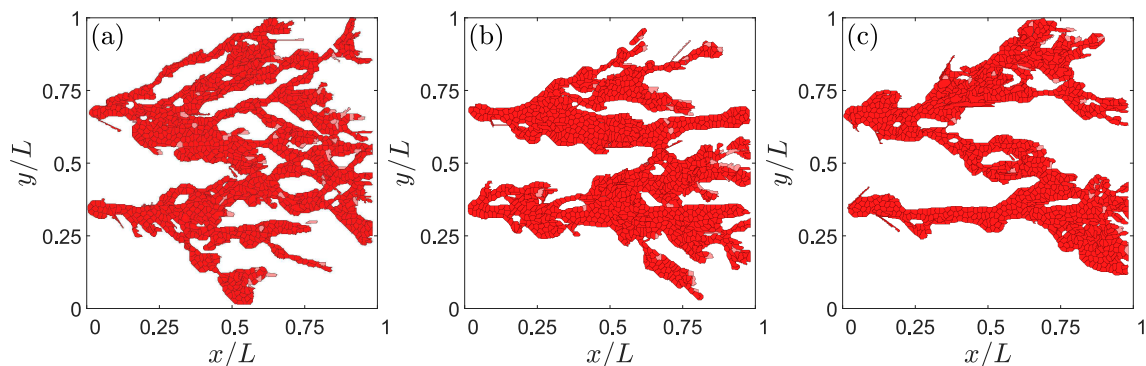
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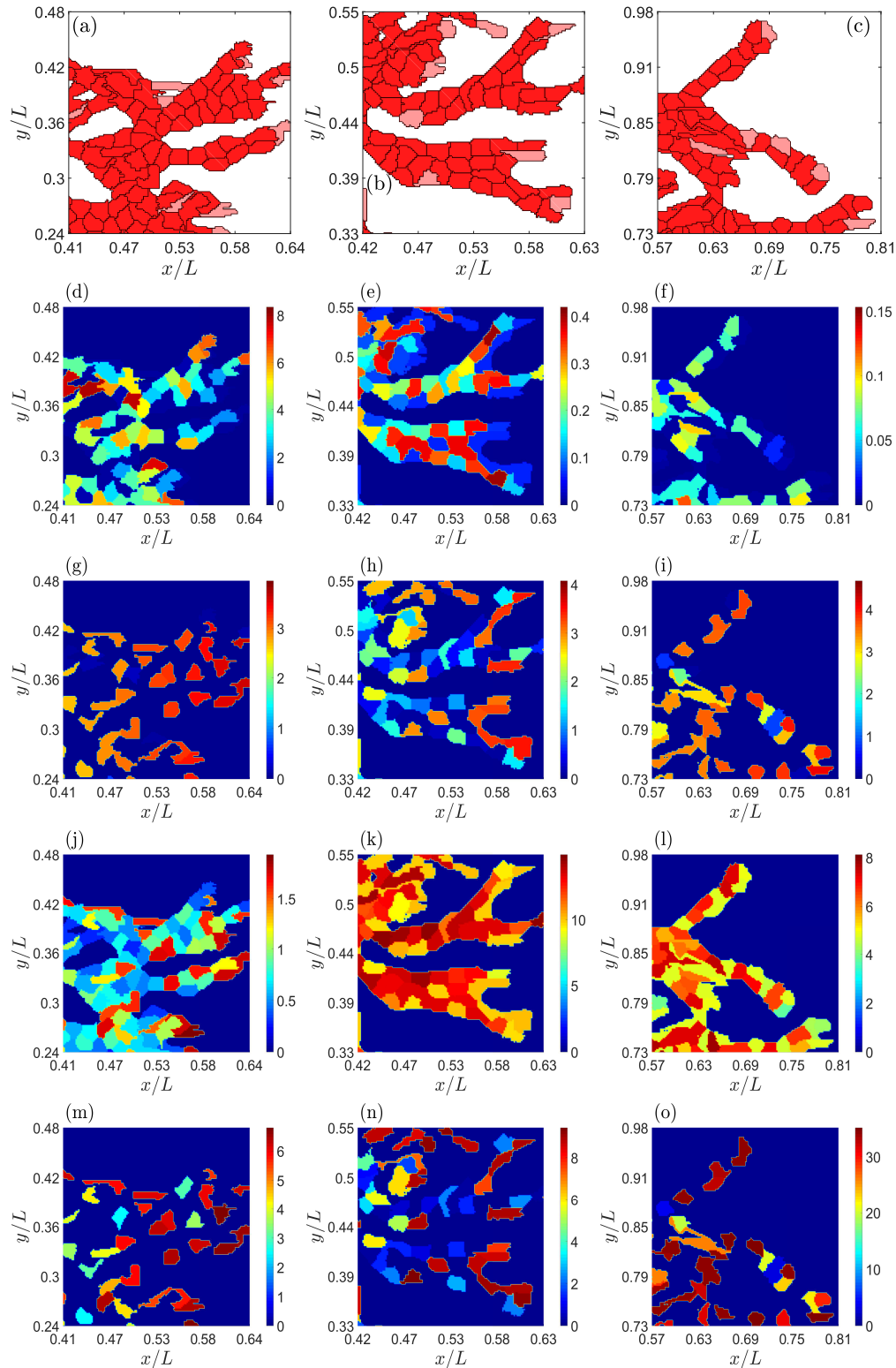
**Figure 3.** Effect of Jagged production on angiogenesis. For  $r_J = 500$  molec/h and  $r_D = 1000$  molec/h, snapshots at times: (a) 2001 MCTS, (b) 2751 MCTS, (c) 3501 MCTS. For  $r_J = 2000$  molec/h and  $r_D = 1000$  molec/h, snapshots at times: (d) 2001 MCTS, (e) 2751 MCTS, (f) 3501 MCTS. For  $r_J = 2000$  molec/h and  $r_D = 7500$  molec/h, snapshots at times: (g) 2001 MCTS, (h) 2501 MCTS, (i) 3501 MCTS.



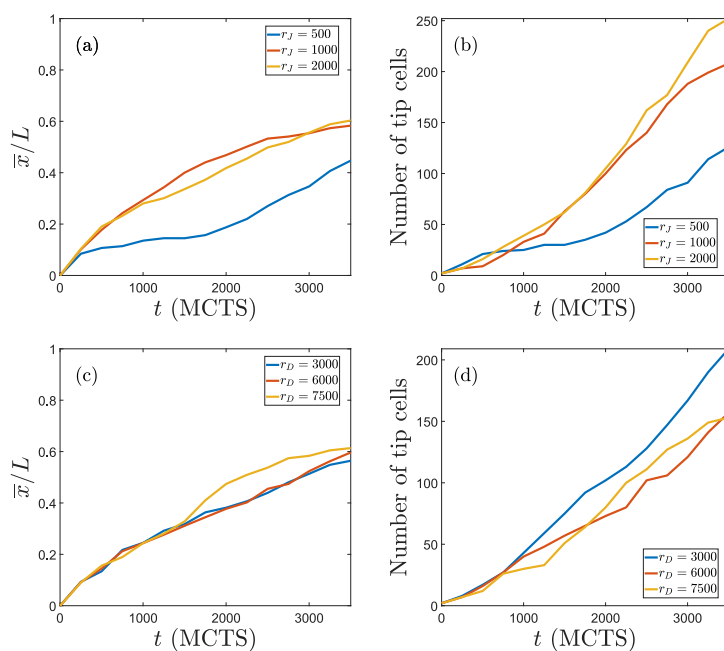
**Figure 4.** Content of VEGF,  $V$ , versus  $V_{\text{ext}} = C$ , and of  $V$  versus  $J$  in the tip, stalk and hybrid tip-stalk cells within the angiogenic network at 3501 MCTS. (a), (b)  $r_J = 500$  molec/h,  $r_D = 1000$  molec/h; (c), (d)  $r_J = 2000$  molec/h,  $r_D = 1000$  molec/h; (e), (f):  $r_J = 2000$  molec/h,  $r_D = 7500$  molec/h. Other parameter values are as indicated in Tables 1-3. Nondimensional units for  $V$ ,  $V_{\text{ext}}$ ,  $J$  are as indicated in Table 4. The meaning of symbols is as follows. Red cross (tip cell), magenta rhombus (hybrid tip/stalk cell), blue circle (stalk cell).



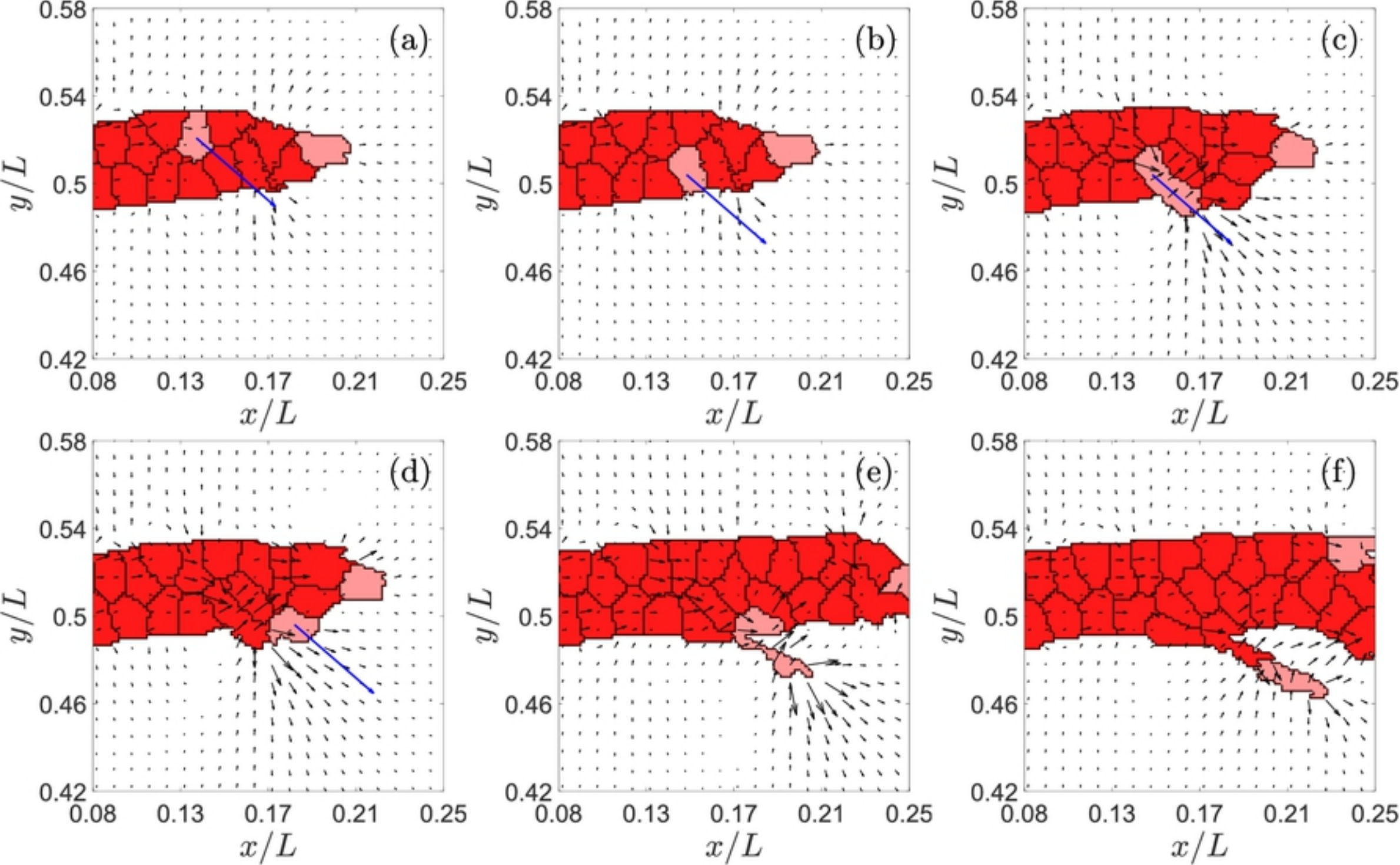
**Figure 5.** Effect of the Delta production rate on angiogenesis with a high Jagged production rate of  $r_J = 2000$  molec/h at 3501 MCTS. (a)  $r_D = 3000$  molec/h, (b)  $r_D = 6000$  molec/h, and (c)  $r_D = 7500$  molec/h. Lateral inhibition due to more activated D-N signaling decreases the number of hybrid tip/stalk cells and branching.



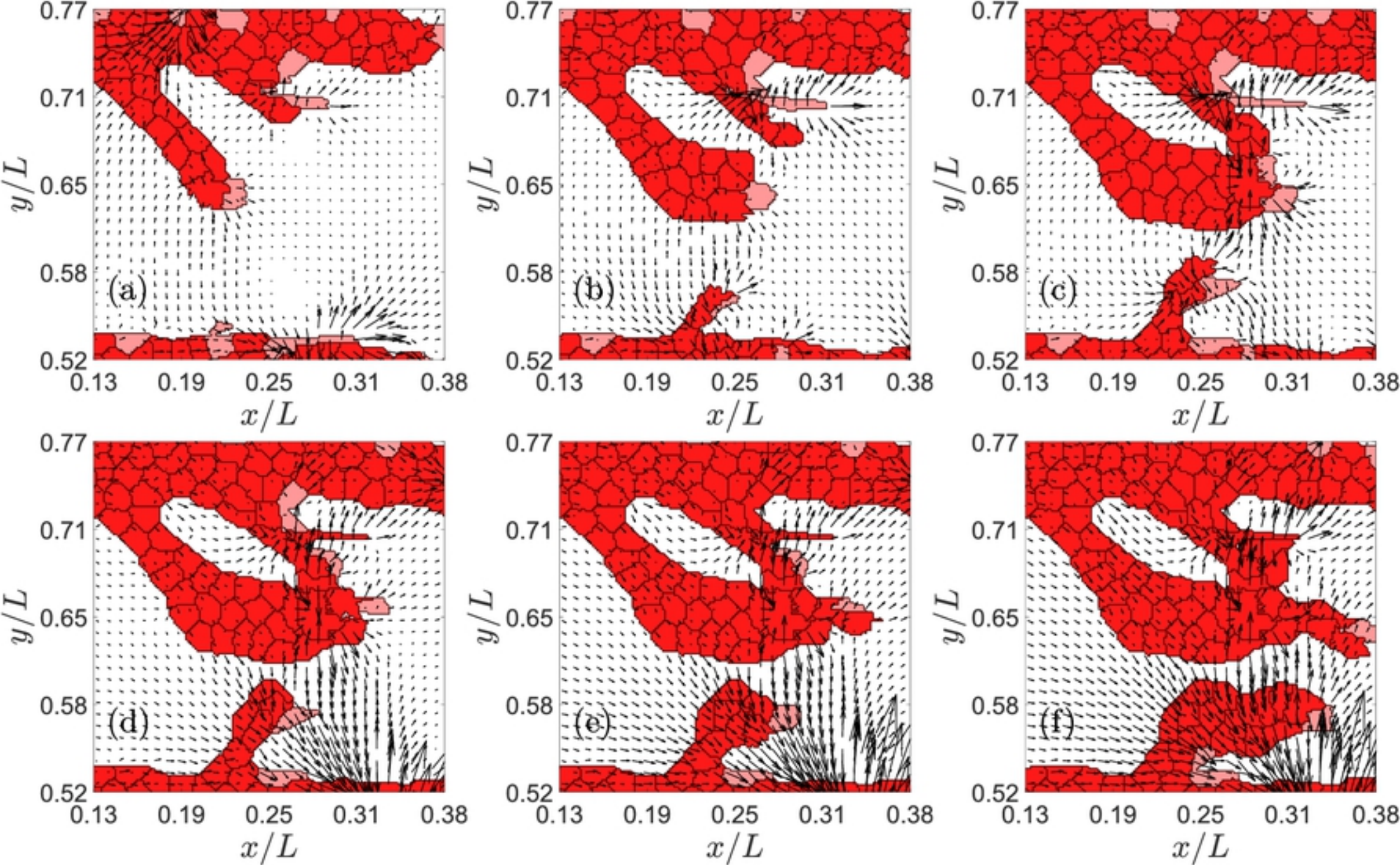
**Figure 6.** Effect of the Jagged and Delta production rates on angiogenesis at a time of 2901 MCTS. (a)-(c) Snapshots of networks, (d)-(f) Notch concentration, (g)-(i) VEGF concentration, (j)-(l) Jagged-1 concentration, (m)-(o) Delta-4 concentration. Data: (a),(d),(g),(j),(m)  $r_J = 500$  molec/h,  $r_D = 1000$  molec/h, (b),(e),(h),(k),(n):  $r_J = 2000$  molec/h,  $r_D = 1000$  molec/h, (c),(f),(i),(l),(o):  $r_J = 2000$  molec/h,  $r_D = 7500$  molec/h. Nondimensional units for protein concentrations are as in Table 4.



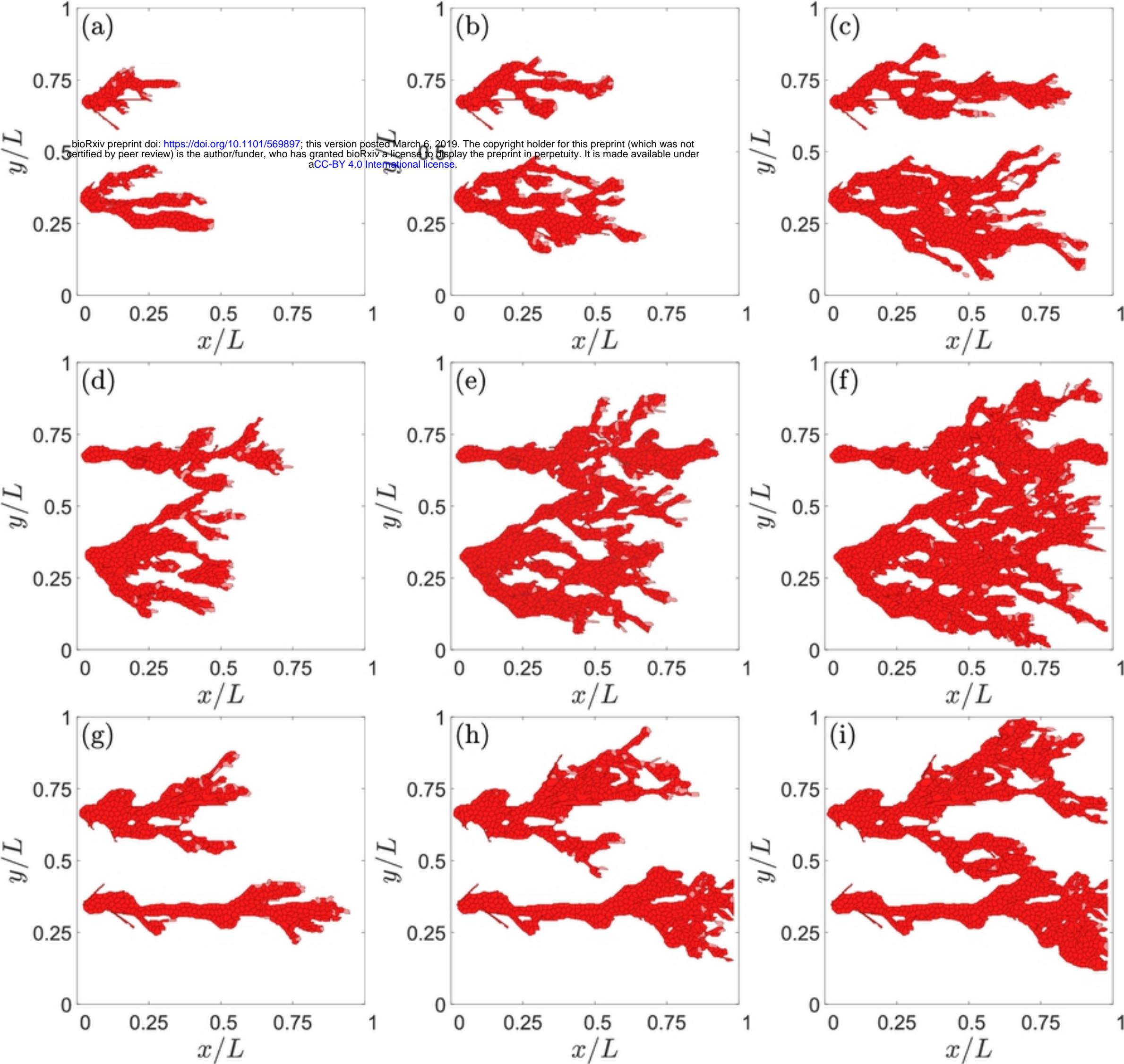
**Figure 7.** (a) Average abscissa (position on  $x$  axis) of the tip cells as a function of time, and (b) number of tip cells versus time for  $r_D = 1000$  molec/h and  $r_J = 500, 1000$  and  $2000$  molec/h. Increasing Jagged production rate yields more tip cells that advance faster. (c) Average position of tip cells versus time, and (d) number of tip cells versus time, for  $r_J = 2000$  molec/h and  $r_D = 3000, 6000$  and  $7500$  molec/h. Increasing Delta production rate makes tip cells to advance slightly more but it diminishes the number of tip cells. The effect of  $r_D$  on the number of tip cells is opposite to that of  $r_J$  in Panels (a) and (b).



Figure

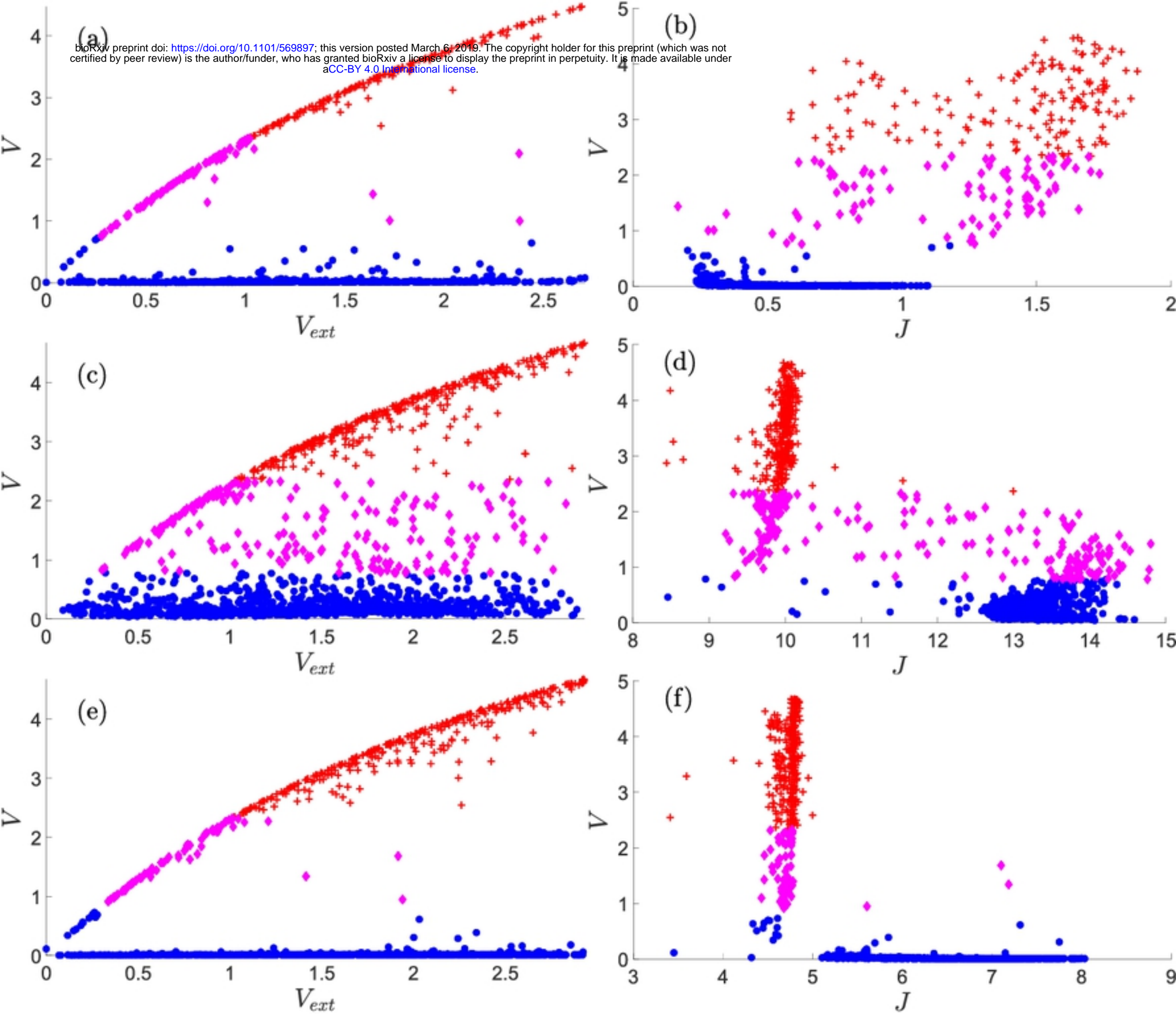


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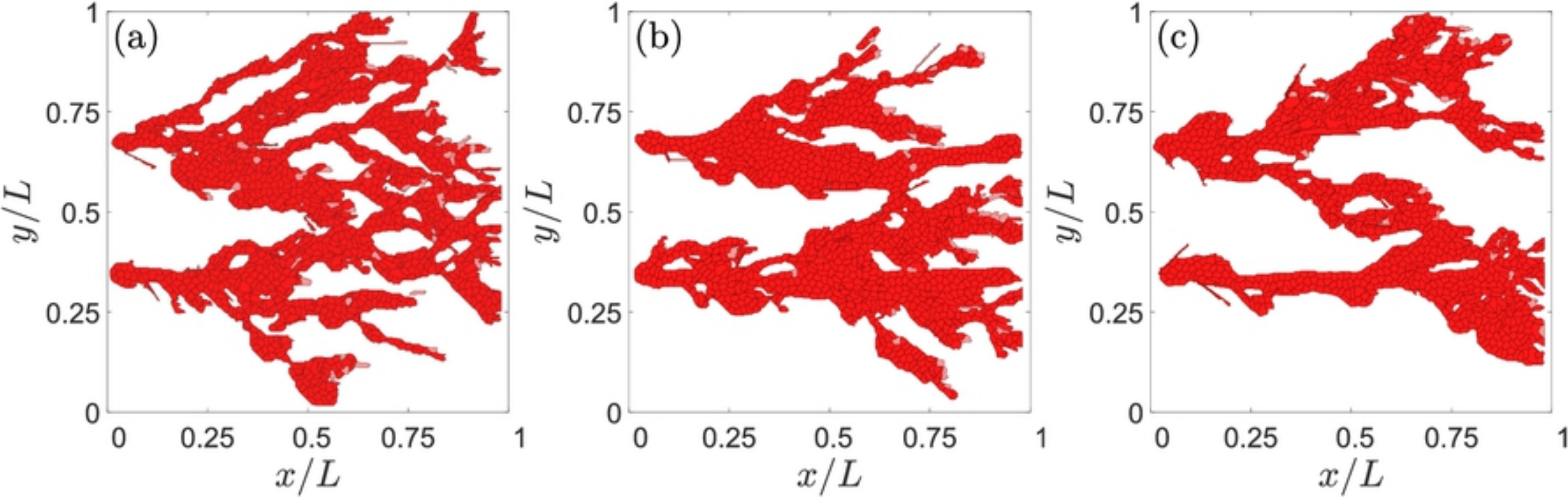


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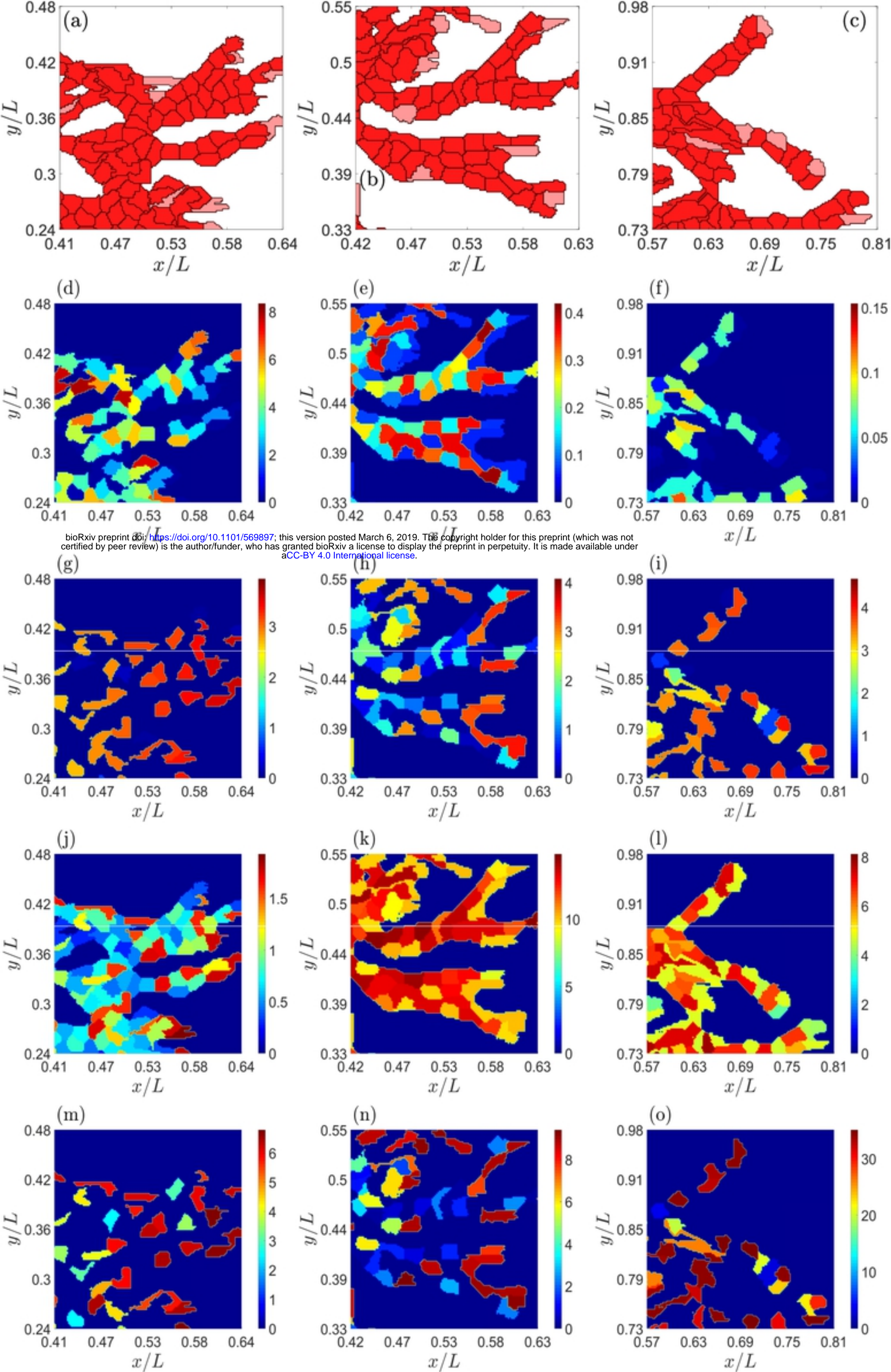




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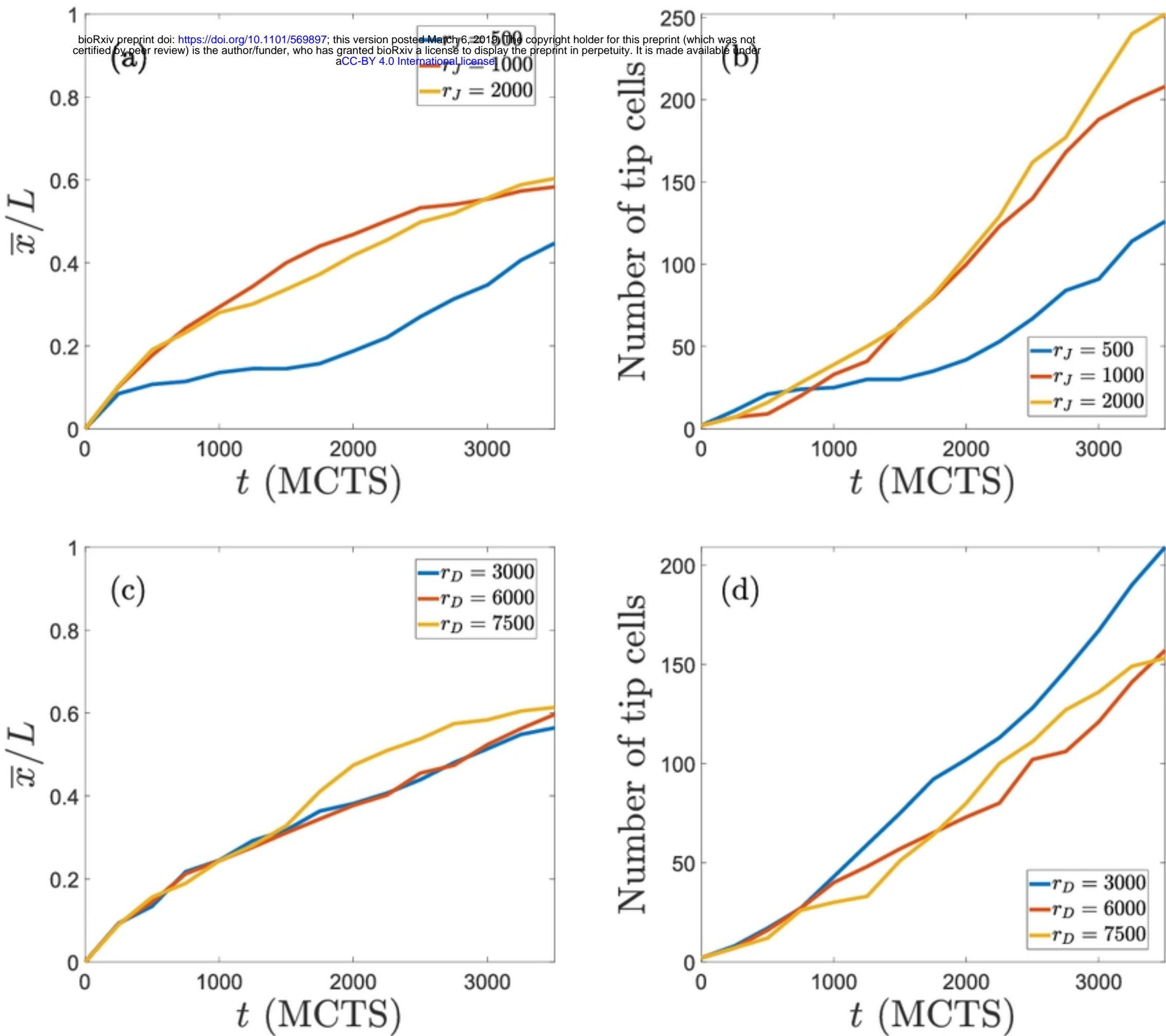


Figure



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Figure



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