Interstitial Flow in Cancerous Tissue: Effect of Considering Remodeled Capillary Network

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Abstract

Interstitial fluid flow has been studied by many cancer researchers for investigation of how interstitial flow can affect drug delivery. This paper couples blood flow through a capillary network induced by tumor angiogenesis and fluid flow in interstitium and investigates the effects of consideration of a remodeled network and adaptability of capillary on interstitial flow. Two approaches are studied for showing the remodeling effects. In the first approach, the capillary network is presumed to have capillary with constant diameter. For the second approach, the adaptable capillaries are assumed in which the metabolic and hemodynamic stimuli changes capillary’s diameter based on sending signals to capillary. Numerical results show that the intravascular pressure has more uniform distribution along the network and capillary diameter has greater value in the network. The prediction of interstitial pressure for the second approach shows higher values compared to the first approach. Results from interstitial flow and blood flow distribution in normal and cancerous tissues have beneficial information about prediction of successive drug delivery.

Keywords: Tumor induced angiogenesis; Capillary network; Remodeling; Blood flow in capillary network; Interstitial fluid flow

Introduction

The abnormal and complicate vasculature induced by solid tumor is introduced as a main reason of unsuccessful drug delivery to tumors. The high interstitial pressure because of high permeability of vessel in tumor region is another barrier to drug transport. The mathematical model used for solid tumor simulation predicts mechanism of drug delivery to tumor cells and helps to find better methods for drug delivery [1,2].

Solid tumor growth modeling involves many complicated processes on a wide range of spatial and temporal scales. The multi-temporal nature of tumor involves formation of vessels induced by angiogenesis which changes in a scale of days and fluid flow and capillary adaptation which occurs in a scale of seconds. The mathematical model of tumor consists of blood flow convection in micro vessel at the micro scale to interstitial fluid flow which includes the convection and diffusion mechanism at the largest scale [3].

The fluid flow modeling in solid tumor as a porous medium by neglecting small scales is common. Soltani et al. [4] introduced the critical tumor radius and critical necrotic radius based on results of fluid flow in tumor done by developed mathematical model of interstitial fluid flow. The effect of shape and size of tumor in interstitial fluid flow is investigated by Soltani et al. [5]. In these works, the capillary network is substituted by uniform distribution of mass source.

The fluid flow simulation in tumor induced vascular network is investigated in many studies. In some research the blood flow is considered only in a single capillary while in other studies, blood flow through a capillary network is coupled with fluid flow in tumor tissue [6-10]. In the latter case, capillary network is generated by mathematical model or image processing from experimental data [6,7,9,10]. None of the above works however considered the non-continuous behavior of blood through capillaries or adaptability of capillary’s diameter.

Not only chemical responses from tumor growth but also functional needs affect the capillary network structure induced by angiogenesis. The functional needs applied in this study are hemodynamic stimuli implemented by blood flow and metabolic stimuli exerted by tissue, respectively. To simulate the structural adaptation of vascular network, a mathematical model based on experimental observation is used. This mathematical model is developed by Pries et al. [11,12]. They investigated structural adaptation of capillary network by a theoretical method. They set metabolic condition and the wall shear stress induced by blood flow and intravascular pressure as minimal requirements to have realistic blood flow in capillary network. In some studies, the blood flow through capillary network in tumor modeled by considering adaptability of capillary and blood behavior in capillaries [13-17].

In spite of the valuable body of work, the comprehensive model which includes remodeling tumor induced capillary network with interaction by interstitial flow is not considered. In the previous work in our group a mathematical model was developed that couples interstitial fluid flow and intravascular fluid flow at the same time by considering adaptability of capillary’s radius to signals from hemodynamic and metabolic stimuli [3]. In the present work, the simulation method presented in our previous work is developed by adding lymphatic system in the mathematical model of interstitial fluid flow, and statistical analysis [3]. Results of mathematical and statistical analysis of two approaches, capillary network with rigid vessels and capillary network with remodeled structure, are compared. The detail

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of both approaches is introduced in Soltani and Chen [3]. Results show that fluid flow modeling through the network with adaptable capillaries has realistic distribution of vessel radii, uniform blood flow, and elevated intravascular pressure in the whole network. This prediction is lead to higher interstitial pressure levels in remodeled network compared to interstitial pressure in the network with rigid capillaries.

Materials and Methods

Angiogenesis, mathematical model

The present Angiogenesis model is motivated by the tumor-induced angiogenesis model initially proposed by Anderson and Chaplain [18]. This mathematical model predicts the capillary formation by tracking the endothelial cells motion in capillary sprout tips. The detail of rules for sprouting angiogenesis and algorithm of this method is determined in our previous work Anderson and Chaplain and Soltani and Chen [18,19].

Remodeling capillary network

The adaptability of capillary’s radius is simulated by mathematical model introduced by Pries et al. [11,20]. The changes of radius are calculated by Pries et al. [11]:

\[
\Delta R = S_{atm} \cdot R \cdot \Delta t
\]

(1)

\[S_{atm} \text{ includes the influences of the hemodynamic } (S_h) \text{, and a metabolic mechanism } (S_m): \]

\[S_{atm} = S_h + S_m - k_s\]

(2)

Where

\[S_h:] \text{ introduced as hemodynamic stimuli and defined by } \]

\[S_h = \log_{10} \left( \left( \tau_w + \tau_{ref} \right) - k_p \log_{10} \tau_e \right) \]

(3)

\[S_m:] \text{ Metabolic stimuli and defined by } \]

\[S_m = k_m \log_{10} \left( \frac{Q_{ref}}{Q_H} + 1 \right) \]

(4)

Where

\[\tau_w:] \text{ the wall shear stress in capillary induced by flow rate, } \]

\[\tau_{ref}:] \text{ A constant value for preventing singular behavior logarithm function } \]

\[\tau_e:] \text{ The wall shear stress induced by the intravascular pressure, } \]

\[10 - 8.6 \exp \left[ -5000 \left[ \log_2 \left( \frac{P_L}{P_H} \right) \right]^4 \right] \]

\[Q_{ref}:] \text{ The largest value of } Q \text{ in the network } \]

\[k_s:] \text{ Shrinking stimuli } \]

\[k_p, k_m:] \text{ Positive real numbers } \]

The detail of these stimulus and their effects are mentioned in our previous work [3].

Intravascular pressure calculation

The blood flow through micro vessel network is modeled similar to electrical network. Since blood flow rate in micro vessel network is equal to electric current in electrical network, the balance of blood flow in each interconnecting point like c can be implemented same as balance of electric current in electrical networks (Figure 1):

\[
\sum_{k=1}^{N} \beta_k Q_k = 0
\]

(5)

Where

\[N:] \text{ The number of adjacent nodes } \]

\[\beta_k:] '1' \text{ shows connection with adjacent node and '0' shows no connection with adjacent node. } \]

\[Q_c:] \text{ The blood flow rate through each capillary (Figure 2), and calculated as } \]

\[Q_c = Q_{bc,c} - Q_{tc,c} \]

(6)

Where

\[Q_{bc,c}:] \text{ Blood flow through each capillary, } \]

\[Q_{tc,c}:] \text{ The flow rate across the microvessel wall or transvascular flow. For blood flow in capillary tubes, Poiseuille’s law can be applied: } \]

\[Q_{bc,c} = \frac{\pi \Delta P_c R_c^4}{8 L \mu} \]

(7)

Where

\[R_c:] \text{ the vessel radius, } \]

\[L:] \text{ the vessel length, } \]

![Figure 1: Schematic of capillary network in a sample node.](image1)

![Figure 2: Schematic convection and extravasation flow through a capillary [3].](image2)
μ: blood viscosity,

$$\Delta P_B$$ calculated by $$P_B^c - P_B^i$$, and $$P_B^c$$ and $$P_B^i$$ are corresponding blood pressure at each node.

The transvascular flow rate is calculated by Starling’s law [21,22]:

$$Q_{tv} = 2 \pi R L_p (P_B^c - P_B^i - \sigma_s (\pi_B^c - \pi_B^i))$$  \hspace{1cm} (8)

where $$P_B^c$$, $$P_B^i$$, $$\pi_s$$, $$\pi_B^c$$, $$L_p$$, and $$\sigma$$ are the interstitial fluid pressure, blood pressure in capillary, capillary oncotic pressure, interstitial oncotic pressure, hydraulic conductivity, and osmotic reflection coefficient, respectively.

The fluid flow governing equation for a porous medium is used to calculate the interstitial pressure for peripheral tissue of a vascular network. The continuity equation for incompressible fluid flow in the porous media is [4]:

$$\nabla \cdot \mathbf{v}_I = \phi_b - \phi_L$$  \hspace{1cm} (9)

where

$$\mathbf{v}_I$$ the interstitial fluid velocity,

$$\phi_b$$ the flow rate from vessel to interstitium or vice versa,

$$\phi_L$$ the flow rate from interstitium to lymphatic system.

The starling’s law represents the fluid flow rate across the microvessel wall [4].

$$\phi_b (r) = \frac{L_p S_v}{V} (P_B^c - P_B^i - \sigma_s (\pi_B^c - \pi_B^i))$$  \hspace{1cm} (10)

The lymphatic drainage term is:

$$\phi_L (r) = \frac{L_p S_L}{V} (P_B^c - P_L^i)$$  \hspace{1cm} (11)

where

$$P_L^i$$: The lymphatic pressure.

$$\phi_L$$ is considered only in normal tissue and in tumor tissue its value is zero [23].

The general form of momentum equation for porous media is as follows [24]:

$$\left( \frac{\mu}{K} \nabla \mathbf{v} \right) - \nabla \left[ -\rho \mathbf{v} + \nabla \left( \mu \mathbf{v} \right) \right] - \frac{2\mu}{3} \nabla \left( \nabla \mathbf{v} \right) - \left( \frac{\mu}{K} \right) \mathbf{v} + \mathbf{F}$$  \hspace{1cm} (12)

where

$$K$$: the permeability of the porous medium,

$$\rho$$: the density of interstitium media,

$$\mu$$: the viscosity of interstitium media,

$$F$$: influence of gravity and other volume forces.

By considering interstitial flow as Newtonian fluid and low velocity for fluid flow and neglecting the friction, equation (12) is simplified to Darcy’s law:

$$\mathbf{v}_p = -\left( \frac{\mu}{K} \right) \mathbf{v}$$  \hspace{1cm} (13)

The $$K/\mu$$ is defined as interstitium hydraulic conductivity $$\kappa$$.

By coupling Equations (9) and (13), considering source term wherever there is a capillary and uniform distribution of sink term, the governing equation for interstitial flow is obtained as follows (For more detail see [3]):

$$-\mathbf{v}_p \cdot \mathbf{F} = \begin{cases} \frac{L_p S_v}{V} (P_B^c - P_B^i - \sigma_s (\pi_B^c - \pi_B^i)) & \text{existence of capillary} \\ \frac{L_p S_L}{V} (P_B^c - P_L^i) & \text{normal tissue} \\ \frac{L_p S_L}{V} (P_B^c - P_B^i - \sigma_s (\pi_B^c - \pi_B^i)) & \text{existence of capillary} \\ 0 & \text{tumor tissue} \end{cases}$$  \hspace{1cm} (14)

**Blood behavior in microvessel**

The dependency of blood viscosity in microvessel to radius and hematocrit introduced as Fahraeus–Lindqvist effect and independency of hematocrit fraction into daughter vessels to blood flow fraction are known as non-continuous behavior of blood [25]. The dependency of viscosity to radius is shown in Figure 3. The detail of non-continuous behavior modeling is mentioned in our previous work [3].

**Calculation algorithm**

Intravascular ($$P_B^c$$) and Interstitial ($$P_B^i$$) pressures are coupled to each other based on Equation (8). Therefore, Equations (5) and (14) should be solved at the same time to find the blood pressure in capillary
network and interstitial fluid pressure. The numerical method is used for solution these equations. The advanced mathematical method is applied to discretize the interstitial fluid flow equation (equation 14) [4].

The general algorithm used in this work is shown in Figure 4. In the first approach, rigid network, the hematocrit, viscosity, and vessel radius are assumed to be constant during simulation. For the second approach, the mentioned parameters are assumed to update during simulation. The detail of numerical procedure is described in Soltani and Chen [3].

Boundary and initial condition and model parameterization

The 2D computational domain (Figure 5), in which parent vessel is located in the left and tumor in the right, is considered for simulation. The radius of circular tumor is assumed 0.4 of domain length.

Boundary and initial conditions

Intravascular flow: To carry out the intravascular flow simulation, pressure values for the inlet and outlet of parent vessel are required. Taking into account the physiological condition at the microvessel scale, the 3325 Pa (25 mmHg) for inlet and 1330 Pa (10 mmHg) for outlet pressures is considered. The initial value of $P_n$ is considered to be 1330 Pa.

The initial radius of each capillary segment for simulation with rigid network is assumed 0.4 of domain length. For simulation with rigid capillary, the vessel radius is prescribed with a method presented by Wu et al. [9] as follows:

$$R_n = g_n R_0 \quad (n=1, 2, 3...)$$

Where $n$ is the value of branching generation, $R_0$ is the radius of the parent vessels, $R_n$ is the radius of the $n^{th}$ generation of vessels, $g_n$ is the ratio coefficient smaller than 1. $g_n$ in equation (15), given by Wu et al. [9]

$$g_{n+1} = \frac{1}{\kappa_{n+1}} g_n$$

In which $\kappa_{n+1}$ is 0.65 and $g_n$ is 0.8. The radius of vessels is remained constant during simulation.

Interstitial flow: The boundary condition used in solving of Equation (14) is shown in Figure 5. Anon-flux (symmetry) boundary is applied in right, down and upper edges, i.e.,

$$VP = 0$$

For the left boundary, the interstitial pressure is constant; the Dirichlet type of boundary condition is applied [4]:

$$P_i = 0 \text{ for outer region}$$

The continuity of pressure and velocity for Equation (14) are considered as appropriate boundary conditions for boundary between normal and tumor tissue:

$$-\kappa_n \nabla P_i |_{\Omega} = -\kappa_n \nabla P_i |_{\Omega'},$$

$$P_i |_{\Omega} = P_i |_{\Omega'}$$

Where $\Omega$ and $\Omega'$ indicate the tumor and normal tissue at the outer edge of the solid tumor; the $t$ and $n$ are the normal and tumor tissue, respectively.

Model parameters: The material properties for normal and tumor tissue listed in Table 1 and used in interstitial fluid flow calculation are taken from our previous work [4]. The parameters used for blood flow simulation through microvessel network are listed in Table 2.

Results and Discussion

Two methods are used for simulation of blood flow in capillary network and peripheral tissues. Results of the first method are obtained by simulating fluid flow in a network induced by tumor angiogenesis with constant values for blood rheology and vessel radius. In the second method, results are obtained using the proposed realistic model assuming the Fahraeus–Lindqvist effect of blood viscosity and remodeling of the capillary network.

The capillary network used in this paper is generated by discrete angiogenesis method with considering 6 endothelial cells in parent vessel. The capillary network shown in Figure 6 (blue line) is used in numerical simulation of intravascular and interstitial pressure. The pruned networks are illustrated in Figure 6 with green lines. Resulted networks agree well qualitatively with the experiments [27,28].

For a better comparison of the results of two approaches, the interstitial pressure is non-dimensionalized. The non-dimensional interstitial pressure is defined by

$$P_i' = \frac{P_i}{P_{i\text{max}}}$$

Table 1: Normal and tumor tissue transport properties [4].

| Parameter | Normal Tissue | Tumor Tissue |
|-----------|---------------|--------------|
| $L_0$ [cm/mmHg s] | $3.86 \times 10^{-3}$ | $2.80 \times 10^{-3}$ |
| $K$ [cm/mmHg s]  | $8.53 \times 10^{-9}$ | $4.13 \times 10^{-8}$ |
| $S/V$ [m^-1] | 700 | 15 |
| $\pi_0$ [mmHg] | 20 | 15 |
| $\pi_i$ [mmHg] | 10 | 0.82 |
| $P_i$ [mmHg] | 0 | - |
| $L_{m} S/V [1/mmHg s]$ | $1.33 \times 10^{-1}$ | - |

Table 2: Adaptation parameter used in remodeling simulation [35].

| Parameter | Value |
|-----------|-------|
| $t_{sa}$ [Pa] | 0.103 |
| $Q_{sa}$ [mm^3/s] | $4.87 \times 10^{-3}$ |
| $k_p$ [1/s] | 0.1 |
| $k_n$ [1/s] | 0.07 |
| $k_m$ [1/s] | 0.35 |
In which $P_{i,\text{max}}$ is the maximum interstitial pressure calculated for two approaches. The calculation shows that the maximum pressure belongs to the remodeled network which is around 1100 Pa. Also, the x and y positions are non-dimensionalized by length of the domain. The 3D graphs of interstitial pressure for two approaches are shown in Figures 7 and 8. At the tumor center, the highest value of interstitial pressure is occurred. The interstitial pressure decreased rapidly towards the periphery. The higher vessel permeability and lack of lymphatic system in the tumor region caused the high elevation of interstitial pressure near the tumor. This effect is observed in our previous works Soltani and Chen and Jain et al. [4,5,29], and in the experimental results of Arifin et al. [30], Huber et al. [31] and Heine et al. [32].

The interstitial velocity contour is shown in Figures 9 and 10 for two calculation methods. The maximum value of IFV is close to the boundary between normal and tumor tissue. Since the IFV based on Equation (13) is only depended to IFP gradient and the IFP has uniform value in tumor region, the IFV is zero. The maximum value...
Figures 13 and 14 show the radius distribution in the domain. The statistical analysis of radius distribution is shown in Figures 15 and 16.

The second approach—remodeled network—predicts the highest interstitial pressure approximately 150% of the highest value of interstitial pressure in rigid network. Results presented in the following figures show how considering realistic conditions such as remodeling and blood behavior in capillaries can affect interstitial pressure.

The blood flow distribution for both rigid and remodeled networks is shown in Figures 11 and 12. The comparison of blood flow for two networks shows that blood flow in remodeled network has a uniform distribution than the blood flow in rigid capillaries. The pruning method based on low flow rate is used in blood flow simulation. This method is based on the experimental observation in which the vessels collapse in low flow and shear stress [34]. Because the capillaries far from parent vessel in rigid network receive low blood flow rate, most of them are eliminated in pruning process. However, in remodeled network, most of these pruned capillaries are retained. The considering structural adaptation in modeling process decreases flow resistance through the capillaries, and also prevents shunt fluid path [11]; therefore, vessels far from parent vessel get more blood flow rate.

Figures 13 and 14 show the radius distribution in the domain. The statistical analysis of radius distribution is shown in Figures 15 and 16.
The density of each radius is calculated and shown for two approaches. As shown in these figures, the number of great radius sizes in remodeled network is much more than that of rigid network. The bigger radius in remodeled network decreases resistance flow in the network and makes uniform distribution blood flow as illustrated in Figure 12.

The uniform distribution of blood flow pattern and radius size affect on intravascular pressure. Figures 17 and 18 show the intravascular distribution for two cases. The average intravascular pressure at each constant x in domain for two approaches is shown in Figure 19. Figure 19 shows that the intravascular pressure from the parent vessel to tumor has a more uniform value for remodeled network in comparison to that of rigid network. Also, an intravascular pressure distribution obtained statistically (Figure 20) shows that the remodeled network has a larger density in higher pressure values in comparison to rigid network.

These results show why the remodeled network predicts higher interstitial pressure than rigid network. The elevated intravascular pressure and high leaky microvessels in tumor region increase the transvascular flow rate and subsequently the interstitial pressure as shown in Figures 7 and 21.

Conclusion

This study provides a numerical simulation of interstitial fluid flow in normal and tumor tissues with distribution of intravascular pressure based on tumor-induced vascular network in two situations: a rigid vascular network and remodeled capillary network.

The proposed method couples different simple models to describe processes ranging from flow through porous media, angiogenesis, blood viscosity in capillaries, phase separation, and structural adaptation of microvessel network. Then it is discussed how these simple methods incorporate important simplifications to complex processes and lead to more realistic results.
The adaptability of capillaries in response to hemodynamic and metabolic stimuli which remodels network structure leads to greater capillary radius compared to the rigid capillary network model. The more-uniform blood flow distribution caused by greater capillary capillary radius compared to the rigid capillary network model. This elevated intravascular pressure in immature blood vessels for remodeled network leads to higher interstitial pressure prediction than interstitial pressure prediction for rigid network. This elevated intravascular pressure in immature blood vessels for remodeled network leads to higher interstitial pressure prediction than interstitial pressure prediction for rigid network.

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