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Space-time least-squares finite element method for convection-reaction system with transformed variables

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Abstract

We present a method to solve a convection-reaction system based on a least-squares finite element method (LSFEM). For steady-state computations, issues related to recirculation flow are stated and demonstrated with a simple example. The method can compute concentration profiles in open flow even when the generation term is small. This is the case for estimating hemolysis in blood. Time-dependent flows are computed with the space-time LSFEM discretization. We observe that the computed hemoglobin concentration can become negative in certain regions of the flow; it is a physically unacceptable result. To prevent this, we propose a quadratic transformation of variables. The transformed governing equation can be solved in a straightforward way by LSFEM with no sign of unphysical behavior. The effect of localized high shear on blood damage is shown in a circular Couette-flow-with-blade configuration, and a physiological condition is tested in an arterial graft flow.

Keywords

convection-reaction system; variable transformation; least-squares finite element method; space-time formulation; nonlinear dynamics; blood damage estimation

1. Introduction

Convection-reaction equations are hyperbolic partial differential equations that arise from various engineering applications, e.g. blood damage estimation in medical devices, petroleum reservoir simulation, and ground-water contamination remediation [1, 2].

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The equations can be derived from the general transport equations (convection-diffusion-reaction equations) by omitting diffusion when the convective flux dominates over diffusive flux. The solutions to these equations usually have moving steep fronts of the transported quantity that lead to serious numerical difficulties. In practice, the computed approximate solutions suffer from (a series of) overshoots and/or undershoots in the quantity; a situation that may cause negative values when only positive ones are expected.

The literature on the computational treatment of convection-reaction systems is not as extensive as that for convection-diffusion systems. Ewing *et al.* [2] approximated a solution by spline wavelets in the Eulerian-Lagrangian localized adjoint method (ELLAM). They focused on shock-tracking by adding fine wavelet corrections near steep fronts. However, the Lagrangian framework requires the method of characteristics, and is not trivial in complicated flow conditions with complex flow geometries. Ha and Kim [3] tested various explicit time-integration schemes for a convection-reaction equation, and showed that some integration schemes broke down more readily than others.

Other than the Lagrangian framework, the discontinuous-Galerkin finite element method (DGFEM) has become popular because it can accommodate shocks or discontinuities naturally [4]. Lacasse *et al.* [5] developed a DGFEM for the hemoglobin concentration estimation in blood devices. Accurate results were obtained for simple flow configurations using an adaptive mesh scheme. Yet, the error analysis required for the adaptive scheme is expensive because the error variable for the scheme must be approximated by basis functions higher-order than that of the concentration variable. Furthermore, some results exhibited negative concentrations, which are physically unacceptable.

Other approaches developed for convection-dominated systems are based on flux limiters, which adaptively blend high- and low-order discretizations, e.g., flux-corrected transport (FCT) and total variation diminishing (TVD) [6, 7]. Originally, standard limiters like *minmod* or *superbee* were designed for the 1D convection equation discretized by finite differences on a uniform mesh. Recently, both the FCT and TVD schemes were adapted to finite element methods, e.g., FEM-FCT [8, 9] and FEM-TVD [10]. When the schemes are successfully applied, they can provide sharp resolution of discontinuities, and therefore maintain the positivity of the transported quantity. In both cases, however, mass lumping is mandatory and there is an ambiguity in the choice of the limiter function [11]. Moreover, FEM-FCT suffers from severe convergence problems in the steady-state limit [10].

In summary, previous methods focused on improving accuracy in simple flow configurations. However, they are not easily applicable to a flow system with complex geometries because: (a) the algorithm based on the shock-tracking requires an adaptive addition of wave packets that is not trivial in handling multiple shock fronts [2]; (b) adaptive mesh algorithms are computationally expensive [5]; and (c) the flux limiter based methods require relatively complicated algorithms in the finite element framework [10, 11].

Here we present a numerical method based on a least-squares finite element method (LSFEM) that is simple enough to apply to complex flow systems, yet sufficiently accurate. We also present a transformed convection-reaction equation that overcomes the problem of negative concentration without requiring major modifications to the method itself.

We choose the problem of mechanical blood damage estimation [1, 12] as an example of a convection-reaction system. In this problem, damage is quantified by the amount of plasma-free (PF) hemoglobin, i.e., hemoglobin in blood plasma. Inside medical devices, it is released from the red blood cell distorted under shear flows. This particular example has an unusual coupling between the convection and reaction terms involving the shear rate, such

that a sudden surge in the shear rate leads to a steep hemoglobin concentration gradient; this can cause the computed concentration to be negative, which is unphysical.

2. Scalar convection-reaction system

2.1. Governing equation for a convection-reaction system

The material balance for the PF hemoglobin can be described as a convection-diffusion-reaction equation:

$$\underbrace{\frac{\partial c_{Hb}}{\partial t}}_{\text{accumulation}} = - \nabla \cdot \left(\underbrace{\mathbf{u} c_{Hb}}_{\text{convective flux}} + \underbrace{\mathbf{j}_{c_{Hb}}}_{\text{diffusive flux}} \right) + \underbrace{r_{\Delta Hb}}_{\text{generation}}, \quad (1)$$

where \mathbf{u} is the fluid velocity, $\mathbf{j}_{c_{Hb}}$ is the diffusive hemoglobin flux, and $r_{\Delta Hb}$ is the PF hemoglobin generation rate. Under normal conditions, the PF hemoglobin concentration in blood flow is small; therefore, its diffusive flux is governed by Fick's law:

$$\mathbf{j}_{c_{Hb}} = - D_{c_{Hb}} \nabla c_{Hb}, \quad (2)$$

where $D_{c_{Hb}} = 6 \times 10^{-7} \text{ cm}^2/\text{s}$ is the self-diffusion coefficient for the hemoglobin [13].

For blood flow in typical biomedical devices, the characteristic length and velocity are $L \sim \mathcal{O}(1 \text{ cm})$ and $V \sim \mathcal{O}(10 \text{ cm/s})$, yielding a Péclet number

$$N_{pe} = \frac{LV}{D_{c_{Hb}}} \sim 2 \times 10^7. \quad (3)$$

In other words, diffusion is negligible in typical device flows. Because blood is incompressible, Eq. (1) becomes a convection-reaction equation. When the total (sum of both RBC and plasma) hemoglobin concentration is constant, the PF hemoglobin concentration c_{Hb} can be written in terms of the ratio f_{Hb} between the PF hemoglobin and the total hemoglobin in the blood. For hemolysis in medical devices, the total hemoglobin is practically constant over time. Then the governing equation for the PF hemoglobin can be written as

$$\frac{\partial f_{Hb}}{\partial t} + \mathbf{u} \cdot \nabla f_{Hb} = r_{\Delta Hb}. \quad (4)$$

The hemoglobin generation rate $r_{\Delta Hb}$ requires a constitutive relationship between PF hemoglobin and shear stress. Garon and Farinas [12] and Farinas *et al.* [14] proposed two different linearized blood damage rate equations (without or with saturation effect) based on an experimental correlation by Giersiepen *et al.* [15]:

$$r_{\Delta Hb}^A = A_{Hb} G_f^{m_{Hb}} \quad (\text{without saturation effect}), \text{ or} \quad (5)$$

$$r_{\Delta Hb}^B = A_{Hb} G_f^{m_{Hb}} (1 - f_{Hb}) \quad (\text{with saturation effect}) \quad (6)$$

where $m_{Hb} = 2.416/0.785 = 3.078$, $A_{Hb} = (3.62 \times 10^{-7})^{1/0.785} (\mu_b)^{m_{Hb}}$, μ_b is blood viscosity, and G_f is a scalar shear rate. We use $G_f = 2 \sqrt{-II_S}$, where II_S is the second invariant of the rate of strain $\mathbf{S} = 1/2(\nabla \mathbf{u} + \nabla \mathbf{u}^T)$. Equation (6) has a saturation term $(1 - f_{Hb})$ that limits the maximum f_{Hb} to unity. For small f_{Hb} , Eq. (6) approaches Eq. (5), asymptotically.

2.2. Computation of flows with recirculations

For steady-state computation, the hyperbolic governing equation (4) needs an inlet boundary condition. However, there is no inlet boundary in a recirculating flow, and thus without the saturation effect in $r_{\Delta Hb}$, Eq. (4) is ill-posed: f_{Hb} grows without bound along recirculating streamlines. With the saturation effect in $r_{\Delta Hb}$, i.e. Eq. (6), the problem affords a sensible steady-state solution, i.e., $f_{Hb} = 1$ in a recirculating flow.

Garon and Farinas [12] proposed a “fast” hemolysis estimation for steady-state analysis based on a global balance. However, the problem mentioned above complicates a global material balance for a steady flow with recirculating zones. According to Garon and Farinas [12], the balance in a sudden-expansion flow (Fig. 1) is,

$$\int_{\Omega} \nabla \cdot (\mathbf{u} f_{Hb}) d\Omega - \int_{\Omega} r_{\Delta Hb} d\Omega = 0, \quad (7)$$

where $\Omega = \Omega_o \cup \Omega_r$. Here, Ω_o stands for open flow domain (with inlet boundary Γ_i and outlet boundaries Γ_o) and Ω_r for recirculating flow domain, bounded by separating streamlines.

After applying the divergence theorem with zero inlet f_{Hb} , the balance equation becomes

$$\underbrace{\int_{\Gamma_o} (\mathbf{n} \cdot \mathbf{u}) f_{Hb} d\Gamma}_{\text{net PF hemoglobin flux}} - \underbrace{\int_{\Omega} r_{\Delta Hb} d\Omega}_{\text{generation inside } \Omega} = 0, \quad (8)$$

$$j_{Hb,o} = \int_{\Gamma_o} (\mathbf{n} \cdot \mathbf{u}) f_{Hb} d\Gamma = \int_{\Omega} r_{\Delta Hb} d\Omega,$$

where Γ_o is the outlet boundary and $j_{Hb,o}$ is the hemoglobin flux across Γ_o . Garon and Farinas [12] used Eq. (8) to estimate $j_{Hb,o}$ by integrating the reaction term over a given flow domain.

The total hemoglobin generation rate inside Ω can be divided into contributions from Ω_r and Ω_o . Therefore, the flux-averaged PF hemoglobin fraction \bar{f}_{Hb} can be expressed as

$$\bar{f}_{Hb} = \frac{j_{Hb,o}}{q} = \frac{1}{q} \int_{\Gamma_o} (\mathbf{n} \cdot \mathbf{u}) f_{Hb} d\Gamma = \frac{1}{q} \left(\underbrace{\int_{\Omega_r} r_{\Delta Hb} d\Omega}_{(A)} + \underbrace{\int_{\Omega_o} r_{\Delta Hb} d\Omega}_{(B)} \right), \quad (9)$$

where $q = \int_{\Gamma} \mathbf{n} \cdot \mathbf{u} \, d\Gamma$ is the flow rate, and (A) and (B) stand for the total hemoglobin generation rate contribution from Ω_r and Ω_o , respectively. The terms (A) and (B) come from two different material balances for Ω_r and Ω_o :

$$\int_{\Omega_r} \nabla \cdot (\mathbf{u} f_{Hb}) \, d\Omega - \int_{\Omega_r} r_{\Delta Hb} \, d\Omega = 0 \quad \text{for } \Omega_r, \tag{10}$$

$$\int_{\Omega_o} \nabla \cdot (\mathbf{u} f_{Hb}) \, d\Omega - \int_{\Omega_o} r_{\Delta Hb} \, d\Omega = 0 \quad \text{for } \Omega_o. \tag{11}$$

The first term of Eq. (10) vanishes, because there is no flux across Ω_r .

For the reaction rate equation without a saturation effect, Eq. (5), the balance equation (10) is not valid because a steady state cannot be achieved in Ω_r . Therefore, Eq. (9) will overestimate the hemoglobin flux by (A).

For the reaction rate equation with a saturation effect, Eq. (6), the balance equation becomes trivial: both terms in the left hand side of the balance equation vanish at the solution $f_{Hb} = 1$. In order to estimate the flux-averaged hemoglobin fraction, Farinas *et al.* [14] proposed a method to evaluate \bar{f}_{Hb} dividing the governing equation by $1 - f_{Hb}$, i.e.

$$\int_{\Omega_o} \frac{\mathbf{u} \cdot \nabla f_{Hb}}{1 - f_{Hb}} \, d\Omega = \int_{\Omega_o} \frac{r_{\Delta Hb}}{1 - f_{Hb}} \, d\Omega = \int_{\Omega_o} r_{\Delta Hb}^l \, d\Omega. \tag{12}$$

Note that the R.H.S. of the equation does not contain f_{Hb} so that the flux-averaged \bar{f}_{Hb} can be evaluated without knowing f_{Hb} inside Ω (See Farinas *et al.* [14] for details). The method cannot, however, be applied in Ω_r , because it requires $f_{Hb} \neq 1$.

In a complex flow, it is not trivial to evaluate the global balance for the open flow domain(s) only. For example, one may need to determine whether a given element belongs to the open flow domain or not. Furthermore, some elements may straddle both open and closed flow domains at the same time. Finally, using continuous basis functions will induce some diffusion across separating streamlines. For these reasons, the concentration field must be computed in the whole flow domain to estimate the outlet flux.

3. Numerical method

3.1. Quadratic variable transformation

In the hemolysis estimation, the reaction term depends on a scalar fluid shear rate G_f from the pre-computed velocity field that is also used in evaluating the convection term. This coupling between convection and reaction is unusual. Moreover, the reaction term is always positive (there is no consumption term and $G_f > 0$).

However, small negative f_{Hb} values were reported, e.g. Fig. 11a of Lacasse *et al.* [5]. The values are typically observed near the regions where G_f changes abruptly, but they are not confined to the regions. They are spread out widely, usually along flow directions most likely due to convection. Overall, this symptom is called a “negative concentration” problem.

The velocity field is represented by continuous linear finite element basis functions so that the velocity gradient (and the shear rate) is discontinuous across element boundaries. The resulting discontinuous reaction term may lead to instabilities in f_{Hb} (caused by sharp gradients of the term [5]). To prevent this, we project the computed velocity gradients onto continuous linear basis functions by a least-squares projection method [16]. However, the projection method was not able to solve the negative concentration problem completely.

We propose a method to overcome the problem by expressing f_{Hb} :

$$f_{Hb} = c^2 + f_{Hb}^b, \tag{13}$$

where c is a real number and f_{Hb}^b is a base value for the PF hemoglobin fraction. Because c^2 is positive, f_{Hb} is always larger than f_{Hb}^b .

For the steady-state computation, one can use the inlet fraction as f_{Hb}^b and set $c = 0$ at the inlet boundary. For the transient problem, f_{Hb}^b can be used as an initial value, with $c^2 = f_{Hb}^{in} - f_{Hb}^b$ at the inlet boundary.

In typical biomedical devices, changes in f_{Hb} between the inlet and outlet of medical devices are lower than order 10^{-6} [12, 17], so that it is desirable to adjust c by a scaling factor κ . Defining $C = \kappa c$, the governing equation (4) becomes

$$\frac{\partial C^2}{\partial t} + \mathbf{u} \cdot \nabla C^2 - \kappa^2 r_{\Delta Hb} = 0. \tag{14}$$

This leads to two possible governing equations:

$$C \left(\frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C \right) - \frac{\kappa^2 r_{\Delta Hb}}{2} = 0, \tag{15}$$

$$\frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C - \frac{\kappa^2 r_{\Delta Hb}}{2C} = 0, \quad (C \neq 0). \tag{16}$$

From now, we will call Eqs. (4), (15), and (16) as the NF, S1, and S2 formulations, respectively.

The S1 formulation may suffer from inaccuracies for small C . The S2 formulation requires special care due to the singularity at $C = 0$. The condition $C = 0$ cannot be used as either an initial guess for the steady-state problem or an initial condition for the transient problem. One may also find difficulty in solving the equations using the standard Newton's method, because of an extremely large residual value for small C ; this difficulty can be overcome by the relaxed Newton's method (see Sec. 3.2).

$C = 0$ can be, however, imposed at the inlet boundary with the finite element method; C values at numerical quadrature points for evaluating the residual for the governing equation are not on boundary. A typical inlet velocity profile has non-zero shear rate that increases

the value of C slightly; exceptions include, for example, plug flow at the inlet. Therefore, C inside the flow domain is guaranteed to be always positive.

Another approach for solving the S2 formulation is to reduce the numerical singularity by modifying the governing equation. Here, we substitute $1/C$ by $1/(C^4 + \varepsilon_r^4)^{1/4}$, where ε_r is a regularization parameter. The regularized formulation becomes

$$\left\{ \begin{array}{l} \frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C - \frac{A_{Hb} G_f^{m_{Hb}}}{2} \left(\frac{\kappa^2}{(C^4 + \varepsilon_r^4)^{1/4}} \right) = 0 \quad (\text{if } r_{\Delta_{Hb}} = r_{\Delta_{Hb}}^A), \\ \frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C - \frac{A_{Hb} G_f^{m_{Hb}}}{2} \left(\frac{\kappa^2(1 - f_{Hb}^b)}{(C^4 + \varepsilon_r^4)^{1/4}} - C \right) = 0 \quad (\text{if } r_{\Delta_{Hb}} = r_{\Delta_{Hb}}^B). \end{array} \right. \quad (17)$$

The effect of the regularizing of $1/C$ is shown in Fig. 2 by comparing the original equation $f_1 = 1/C$ with the regularized equations with different ε_r : f_2, f_3 , and f_4 . For example, the function with $\varepsilon_r = 0.01$ yields virtually the same value as $1/C$ for $C \gtrsim 0.01$. Note that the fourth power in the regularized equation is chosen such that it leads to a sharp “cut-off” of large value near $C \sim \varepsilon_r$.

For successful solution of both unregularized and regularized S2 formulations, the choice of κ^2 is important. The computed C values must be reasonably far away from $C = 0$ in order to avoid numerical singularity. At the same time, the values need to small enough to avoid the loss of accuracy due to high sensitivity near large C . Note that $f_{Hb} \propto C^2$. With a proper κ^2 , we found that $\varepsilon_r \leq 0.01$ is a reasonable choice.

3.2. Space-time least-squares finite element method

We choose a least-squares finite element method to solve the governing equation. The method can be used in a straightforward manner for any of the formulations presented: NF, S1, and S2. To handle the transient problem, the space-time formulation is deployed; the finite element basis functions for space are continuous, but the basis functions for time are discontinuous across a space-time slab interfaces [18]. The resulting method is called the space-time least-squares finite element method (ST/LSFEM).

There have been several studies on this method, for example Jiang and Carey [19] and Pontaza and Reddy [20]. Here, we introduce a “jump” term in the least-squares functional that weakly imposes continuity of variables across the space-time slab interface. For a homogeneous scalar equation with linear-in-time approximation, the method is known to have third order accuracy [21].

In the space-time formulation, we partition the time interval $[0, T]$ into subintervals $I_n = [t_n, t_{n+1}]$. Letting the space domain at time t_n be denoted $\Omega_n = \Omega_{t_n}$ and its boundary $\Gamma_n = \Gamma_{t_n}$, one can define a space-time slab Q_n as the domain enclosed by the surfaces, Ω_n and Ω_{n+1} , and P_n , where P_n is the surface described by the boundary Γ_n as t traverses I_n . Here, we specify the variable (f_{Hb} or C) at the inlet boundary of the space-time domain $(P_n)_g$. For each space-time slab interface, we define finite element basis function space for the variable:

$$(S^h)_n = \{S^h | S^h \in H^{1h}(Q_n), S^h \doteq g^h \text{ on } (P_n)_g\}. \quad (18)$$

The least-squares statement for a space-time slab Q_n can be written as:

$$\mathcal{F}_n = \int_{\Omega_n} R_M^2(a) dQ + \underbrace{\int_{\Omega_n} ((a)_n^+ - (a)_n^-)^2 d\Omega}_{\text{jump term}} = 0, \quad (19)$$

where the variable a depends on the formulation: $a = f_{Hb}$ for $M = \text{NF}$, and $a = C$ for both $M = \text{S1}$ and $M = \text{S2}$. In the above equation, the following notations are applied:

$$a^\pm = \lim_{\varepsilon \rightarrow 0} a(t_n \pm \varepsilon)$$

$$\int_{\Omega_n} \cdot dQ = \int_{I_n} \int_{\Omega_n} \cdot d\Omega dt.$$

The approximated solution by finite element basis functions is obtained by minimizing the weighted residuals:

$$\mathcal{R}_M^i = \int_{\Omega_n} \frac{\partial R_M(a)}{\partial a^i} R_M(a) dQ + \underbrace{\int_{\Omega_n} (\varphi^i)_n^+ ((a)_n^+ - (a)_n^-) d\Omega}_{\text{jump term}}, \quad (20)$$

where $a = \sum_i^N \varphi^i a^i$. Here, φ^i is a continuous linear basis function and a^i is its coefficient. Tables 1 and 2 respectively summarize the equations for the raw residuals R_M , and the sensitivities of the residuals with respect to variables $\partial R_M / \partial a^i$. Note that the jump term in Eq (20) only appears when node i belongs to Ω_n . Hence, the overall effect of the jump term is minimizing the discontinuity of variable a across Ω_n .

We used Newton’s method for the NF and S1 formulations. However, during Newton’s method for S2 formulation, a sudden surge in the norm of residual vector was observed due to the $1/C$ term; the norm does not decrease below a certain value. A relaxed Newton’s method based on an adaptive relaxation parameter can alleviate this problem.

In the relaxed Newton’s method, N non-linear algebraic equations for a^i (c.f. Eq. (20)) are solved iteratively using:

$$\left. \begin{aligned} \sum_{j=1}^N \mathcal{J}_M^{i,j} \delta a^{j,(m)} &= -\mathcal{R}_M^i \\ a^{j,(m+1)} &= a^{j,(m)} + \rho_r^{(m)} \delta a^{j,(m)} \end{aligned} \right\} \text{ for } i=1, 2, \dots, N, \quad (21)$$

where $\rho_r^{(m)}$ is the relaxation coefficient, and $\mathcal{J}^{i,j}$ is the Jacobian components that have linear terms only, and is defined as:

$$\mathcal{J}_M^{i,j} = \frac{\partial \mathcal{R}_M^i}{\partial a^j} = \int_{\Omega_n} \frac{\partial R_M(a)}{\partial a^i} \frac{\partial R_M(a)}{\partial a^j} dQ + \underbrace{\int_{\Omega_n} \varphi^i \varphi^j d\Omega}_{\text{jump term}}. \quad (22)$$

Here, the superscripts m and $m + 1$ indicate the current and next iteration steps. Note that $\rho_r^{(m)} = 1$ corresponds to the standard Newton’s method (used for NF and S1 formulations). We applied the method until the \mathcal{L}_2 norm of residual vector $\mathbf{R}_M^i = [\mathcal{R}_M^1, \mathcal{R}_M^2, \dots, \mathcal{R}_M^N]^T$ or solution

update vector $\delta \mathbf{a}^{(m)} = [\delta a^{1,(m)}, \delta a^{2,(m)}, \dots, \delta a^{N,(m)}]^T$ falls below a certain threshold ε_t , e.g., 10^{-8} .

We use an update scheme for the $m+1^{th}$ step relaxation parameter $\rho_r^{(m+1)}$ based on

$$\rho_r^{(m+1)} = \begin{cases} 0.5 \rho_r^{(m)} & \text{update (A)} \\ 1.5 \rho_r^{(m)} & \text{update (B)} \end{cases} \quad (23)$$

Update (A) occurs when a norm of current step's residual vector is greater than a norm of previous step's residual vector for m_A consecutive steps. Update (B) occurs when a norm of current step's residual vector is less than a norm of previous step's residual vector for m_B consecutive steps. Here, the minimum and maximum allowed $\rho_r^{(m)}$ during update are 0.1 and 1, respectively. We found that $m_A = 1$ and $m_B = 5$ worked successfully for many cases.

For the steady-state formulation, all jump terms in Eqs. (19) – (22) need to be neglected. Integrals over the space-time slab in those equations have to be re-written in terms of integral over the spatial domain only, i.e., substitute $\int_{Q_n} \cdot dQ$ to $\int_{\Omega} \cdot d\Omega$. Also, all derivatives with respect to time in Tables 1 and 2 need to be ignored.

4. Numerical experiments

A major problem in the numerical treatment of a convection-reaction system manifests itself as an unwanted numerical diffusion; it is almost impossible to exclude this diffusion completely. Even with the adaptive discontinuous Galerkin finite element method, some solutions were not free from such diffusion [5].

Another serious problem is caused by the numerical diffusion across separating streamlines. When one of the sides across the streamline has flow recirculation, a large concentration jump across the streamline is possible, and may lead to high artificial diffusion. Here, we test the effect of numerical diffusion across the streamline in the context of ST/LSFEM in a simplified flow system modified from Lacasse *et al.* [5].

4.1. Two-dimensional recirculating flow

The flow is based on the two-dimensional velocity profile $\mathbf{u} = \mathbf{i}u + \mathbf{j}v$:

$$u = \mathcal{A} (2 - 2y) \left(\frac{x^3}{4} - x \right), \quad (24)$$

$$v = \mathcal{A} (2y - y^2) \left(1 - \frac{3x^2}{4} \right), \quad (25)$$

where \mathcal{A} is chosen as 0.133 to make the maximum G_f unity.

The velocity profile sets the separating streamline aligned to y axis as shown in Fig. 3. There are two different types of flow on either side of the separating streamline: the inlet/outlet flow in Ω_{IO} and the recirculation flow in Ω_R . Using this flow system, we will test the numerical diffusion across the streamline for different mesh configurations and formulations (NF, S1, and S2).

To test the effect of the magnitude of the reaction rate on the numerical diffusion, we use a convection-reaction equation in the following form:

$$\frac{\partial n}{\partial t} + \mathbf{u} \cdot \nabla n - r = 0, \quad (26)$$

model reaction:

$$r = A G_f^m d^* (1 - n), \quad (27)$$

where n is the normalized concentration, m is chosen as unity, A is the magnitude factor for the rate r , and d^* is a modification factor defined as

$$d^* = \begin{cases} 1 & \text{if } \|\mathbf{x} - \mathbf{x}_c\| \geq 0.2 \\ \left(\frac{\|\mathbf{x} - \mathbf{x}_c\|}{0.2} \right)^3 & \text{if } \|\mathbf{x} - \mathbf{x}_c\| < 0.2 \end{cases}, \quad (28)$$

which suppresses a spurious peak near the corner \mathbf{x}_c arising from treating the corner as the inlet boundary. Note that the corner \mathbf{x}_c belongs to both inlet and outlet boundaries. Without d^* , either treating \mathbf{x}_c as a part of inlet or outlet leads to a spurious peak near \mathbf{x}_c . We designed and tested d^* such that it does not affect solutions near the separating streamline. Here, we set $n = 0$ for the inlet boundary.

To test the numerical diffusion across the separating streamline, we prepared three different meshes as shown in Fig. 4. The structured mesh SM has the relevant mesh boundaries aligned to the separating streamline, but the unstructured mesh UM does not. A solution free of the diffusion across the streamline was obtained by the special mesh DSM. The mesh DSM is constructed from two distinct structured meshes for Ω_{IO} and Ω_R , which hereafter are respectively called IOSM and RFSM. The solution is obtained by combining results from both meshes after solving on them separately. The effect is the same as using discontinuous basis functions across the streamline in order to capture a discontinuous concentration profile. Note that there are two different concentration variables per node along the separating streamline.

As discussed in Sec. 2.2, the “true” concentration profile is discontinuous across the separating streamline, especially for small A . On the streamline, n is always unity in RFSM for any A , but is different from unity in IOSM for small A (Fig. 5).

Solutions always suffer from numerical diffusion, when the discontinuous concentration profile is not treated by a special method. The degree of this diffusion increases as A decreases because the jump of n across the streamline increases. When mesh boundaries are not aligned to the streamline, the solution has more diffused concentration profile near the streamline (i.e., UM yields more diffusive results than SM). Figures 6 and 7 show the concentration profile along the outlet boundary (bold line) and $y = 0.5$ (dotted line). Solutions on both SM and UM for $A = 1$ capture accurate n profiles. For $A = 0.1$ and 0.01 , n inside the vortex deviates from unity and the deviation increases as A decreases. For these A values, both the recirculation flow and the inlet/outlet flow suffered from numerical diffusion.

When A is decreased further, e.g., $A = 0.001$, n inside the vortex becomes order of 10^{-2} . Unlike for moderately small A ($A = 0.1$ and $A = 0.01$), the concentration profiles within Ω_{IO} on SM and UM become close to the profile in IOSM, as shown in Fig. 7(c) and (d).

We checked the effect of the quadratic variable transformation on the numerical diffusion issue. For the transformed governing equations, we used the scaling factor $\kappa^2 = 10^2$. Figure 8 shows a comparison in the solutions of the original equation (NF) and the transformed ones (S1 and S2). In general, the S2 formulation yields larger n than that of the other formulations inside a flow recirculation zone.

We also checked the effect of mesh type by comparing results from SM and UM. Usually, n values inside the vortex in SM have larger values than those from the other meshes. In Ω_{IO} , n values are virtually the same except near the separating streamlines, regardless of formulation or mesh choice.

There is a small deviation in n profile near the corner \mathbf{x}_c for S1 formulation as shown in the inset of Fig. 8(a): streamlines turn sharply toward the outlet, and their lengths are short. In other words, inaccuracies in concentration do not propagate over a long distance. They may, however, lead to problems in other flows that have long streamlines across regions of low concentration. Based on the above analyses, we decided to use the S2 formulation.

We conclude that ST/LSFEM with transformed governing equations can capture reasonably accurate concentration profiles of the PF hemoglobin, except in the recirculating flow region(s). For hemolysis estimation, A is extremely small (6.2306×10^{-9} , c.f. Eq. (5)). Therefore, in steady-state computations, the method cannot predict accurate concentrations inside regions of flow recirculation, where the concentration is already known without computation. However, the computed field outside flow recirculation are reasonably accurate and so is the outlet hemoglobin flux.

5. Examples

In this study, we consider two examples: a modified circular Couette flow and an arterial graft flow.

5.1. Circular Couette flow with blade configuration

Circular Couette devices have typically been used in blood damage experiments [22], because they offer a simple way to impose a uniform shear rate on the RBCs. A typical device consists of two concentric cylinders, where the outer cylinder rotates at a constant angular speed (setting the shear rate). In the absence of Taylor-Couette vortices, the governing equation (4) can be easily reduced to an ordinary differential equation.

However, when a blade is introduced, the streamlines lose concentricity so that f_{Hb} needs to be computed by solving Eq. (4) directly. The blade imposes a periodic high-shear impulse that is akin to a flow in an axial blood pump.

Here, we consider the two-dimensional flow with following configuration. The inner and outer cylinder radii are $R_i = 34$ mm and $R_o = 35$ mm. The outer cylinder is rotating with a prescribed angular speed $\omega(t) = 10 \times \min(t, 0.5)$ rad/s. The blade is 0.5 mm tall and 4 mm wide, and is attached to the inner cylinder. The corners are rounded with the radius 50 μm to reduce the stress singularity, as shown in Fig. 9(a).

The velocity field of the blood flow (assumed to be Newtonian) was computed by Galerkin/least-squares (GLS) finite element method with bilinear basis functions for both velocity and

pressure (for a review of governing equations for the blood flow and the GLS method, see Behbahani *et al.* [23]).

To illustrate the negative concentration problem, the f_{Hb} field was computed from the NF formulation. The time step size is $\Delta t = 0.01$ s, and the linearized reaction rate with the saturation effect, Eq. (6), is used (the scaling factor $\kappa^2 = 10^7$). Even with a non-zero initial condition (e.g. $f_{Hb,i} = 10^{-13}$), there are regions with negative f_{Hb} values, as shown in Fig. 10(a), where the magnitude of the minimum negative value (-5.1×10^{-9}) is comparable to that of the maximum positive value (2.1×10^{-7}). Some of the negative f_{Hb} values were spread along the flow direction, as shown in regions (A) and (C) of Fig. 10(a), whereas some of the negative values appear near region (C) where the shear rate G_F is changing abruptly with showing an oscillatory behavior (see Fig. 10(b) for the shear rate contour plot). However, it is difficult to find a direct correlation between a negative value and an oscillatory shear rate region. For example, non-oscillatory G_F regions (A), (B), and (D) also show negative f_{Hb} .

To prevent negative concentrations, we use the S2 formulation to compute f_{Hb} . The initial condition was the same as for the NF formulation ($C = 0.001$ with $\kappa^2 = 10^7$). Because f_{Hb} and C in this specific flow system always increase with time, there is no point to use the regularized formulation.

The computed f_{Hb} is converted to a normalized index of hemolysis (NIH), which is widely used in reporting blood damage. The value is directly converted from f_{Hb} according to [17]:

$$\text{NIH}(\text{g}/100 \text{ L blood}) = 100 \times f_{Hb}^{0.785} \times \left(1 - \frac{H_{ct}}{100}\right) \times \kappa_c, \quad (29)$$

where H_{ct} is the blood hematocrit (45% for a healthy person) and κ_c is the hemoglobin content of blood (150 g/L for a healthy person). Note that $f_{Hb}^{0.785}$ is used in evaluating NIH because we use the linearized reaction rate [12, 14].

Figure 11 shows the time evolution of NIH values near the blade (the scale in the plots is set to emphasize the high-NIH regions). The maximum NIH value during the transient computation ($t < 1.0$ s) was $\text{NIH} = 0.0638$ on the blade surface close to the downstream corner. The NIH was found to oscillate slightly inside zone (A) as indicated in Fig. 11, especially for $t > 0.5$ sec. This zone coincides with the downstream region of oscillatory G_F in Fig. 10(b). One can reduce this spurious oscillation by increasing node density near the blade corner or adjusting the mesh to align it with the streamline.

A large vortex is observed downstream of the blade as shown in Fig. 9(b). NIH values inside the vortex increase over time, as shown in Fig. 11(c) to (e). Note that the vortex not only hemolyzes RBC but also traps the PF hemoglobin near the blade. However, the low shear zone (B) slightly below the vortex exhibits low NIH values. This is a clear evidence that there is another separating streamline below the large vortex that prevents an influx of the hemoglobin. Because we did not round bottom corners of the blade near the inner cylinder, there must be a cascading vortex structure near these corners (often called corner vortices) as predicted by Moffatt [24]. However, the mesh is not designed sufficiently to capture such a small flow feature. The lack of precision in the flow solution may lead to an inaccuracy in the NIH prediction in the corners, but the important features of the solution should not be affected, because the effect is locally-confined by separating streamlines. Moreover, the scale of such vortex structures can be comparable to or smaller than the size of RBC. In

short, it may not be suitable to consider the effect of this small flow feature in hemolysis estimation.

The hemoglobin concentration near the blade changes slowly for $t > 0.5$ s, but the generated PF hemoglobin contaminates the entire flow domain by convection. Figure 12 shows the evolution of NIH values for the whole flow system over time. Here the maximum NIH value for the plots was set to 0.001. Clearly, the region of high concentration grows along the rotation direction, as shown from (c) to (e) of Fig. 12.

5.2. Flow in a curved pipe

We also examined a flow in a curved pipe grafted to a straight channel. The flow can be considered as an idealized bypass graft which is used as an alternative conduit around critically stenosed arteries. This particular flow problem was previously subject to the shape optimization of the bypass passage as discussed by Abraham *et al.* [25], who optimized the shape of the bypass graft based on a dissipation function computed from the scalar shear rate G_f . For simplicity, our idealized graft flow assumes that the shape is an arc of circle.

For computing the velocity field, the GLS method is used for both steady-state and transient computations (see Abraham *et al.* [25] for details). We review a dimensionless number and some boundary conditions here.

We define the Reynolds number as

$$N_{Re} = \frac{\rho U_{\max} H}{\mu}, \quad (30)$$

where $\rho = 1,056 \text{ kg/m}^3$, $\mu = 3.5 \text{ cP}$, $H = 0.8 \text{ cm}$, and U_{\max} is the maximum x -velocity at the inlet boundary. We considered $N_{Re} = 300$ (corresponding to a physiological condition) for both steady-state and transient flows.

At steady-state, the flow at the inlet boundary is prescribed to have a parabolic velocity profile $\mathbf{u}(y) = \mathbf{i} U(y)$. For transient flow, the inlet velocity profile is modulated by $m(t)$:

$$\mathbf{u}(y) = \mathbf{i} m(t) U(y), \quad (31)$$

where $m(t)$ is defined as

$$m(t) = \begin{cases} 0.2t & 0 \leq t < 5, \\ 1 & 5 \leq t < 8, \\ 1 + 0.2 \sin[\omega(t - 8)] & 8 \leq t, \end{cases} \quad (32)$$

with $\omega = 2\pi$ corresponding to 60 beats per minute (BPM). The outlet flow is assumed to be fully developed: $\mathbf{n}_o \cdot \nabla \mathbf{u} = 0$, where \mathbf{n}_o is the normal vector pointing outward from the outlet boundary.

Figure 13 shows the geometry of the idealized graft flow. Among the geometric parameters, the length of downstream blood vessel l_4 has to be chosen carefully for two reasons: (a) for the velocity field computation, it must be long enough to ensure a fully developed flow at the outlet especially for the steady-state computation; (b) for the hemolysis estimation, the

vortex needs to be generated far away from the outlet boundary, otherwise the outlet NIH flux will be overestimated for the steady-state computation and will show noise-like temporal fluctuations for the transient computation. We found that $l_4 = 2 \times l_1 = 6$ satisfies both requirements.

Figure 14(a) shows the streamline plots for the steady-state computation. Large recirculations are found near the blocked passage, Area I and Area II. Some recirculations are located downstream of the high shear zones, Vortex 1 and Vortex 4. Note that corners (A) and (C) provide high shear rates as shown in Fig. 14(b).

For both steady-state and transient computations, we set the inlet $f_{Hb} = 0$. To prevent negative concentrations, we use the S2 formulation with $\kappa^2 = 10^9$.

5.2.1. Steady-state results—For the steady-state hemolysis estimation, we tested both the original S2 formulation and the regularized formulation with $\varepsilon_r = 0.05, 0.01, \text{ and } 0.001$. The adaptive relaxed Newton's method was used with $m_A = 1$ and $m_B = 5$ for both formulations. Without the help of relaxation, the \mathcal{L}_2 norm of residual vector cannot decrease below 10^{-8} .

Figures 14(b) and (c) show G_f and NIH value contour plots from the original S2 formulation. The maximum NIH value was 2.7×10^{-3} located at the bottom of the blood vessel near the downstream impact point (D), where G_f has the highest value. The maximum value in the NIH plot was set to 2×10^{-4} for emphasizing high NIH regions. A trail of high NIH appears downstream of the impact point, as shown in Fig. 14(c). The streamline starting from the upstream high shear point (B) passes through the impact point (D) as well. This result suggest that it is important to adjust the entry and exit angles of the graft for decreasing the shear rate at both (B) and (D).

As discussed in Sec. 2.2, the steady-state solution cannot capture $f_{Hb} = 1$ inside flow recirculation regions. Plots (a) and (b) in Fig. 15 present f_{Hb} along the mid plane of the blood vessel (line 1 and line 2 in Fig. 14(b)).

As predicted in Sec. 4.1, The f_{Hb} values in the flow recirculations deviate significantly from unity. This is a major drawback of LSFEM method for the steady-state hemolysis estimation. Nevertheless, the concentration profile outside the vortices does not suffer from numerical diffusion significantly, as already shown in the examples of Sec. 4.

One can neglect f_{Hb} inside the recirculation regions (where $f_{Hb} = 1$) and compute the outlet NIH flux for estimating the blood damage. Here the average outlet NIH flux is defined as

$$\text{NIH}_{\text{avg}} = \frac{\int_{\Gamma_o} \text{NIH} (\mathbf{u} \cdot \mathbf{n}_o) d\Gamma}{\int_{\Gamma_o} (\mathbf{u} \cdot \mathbf{n}_o) d\Gamma}. \quad (33)$$

For the graft, $\text{NIH}_{\text{avg}} = 1.43 \times 10^{-5}$ g/100L blood.

Figure 16 shows the comparison between the NIH profiles along the outlet boundary computed from the original S2 formulation and the regularized one. The plots confirmed that the regularization has virtually no effect on the hemolysis estimation.

5.2.2. Transient results—For transient computations, the time domain $[0, 20 \text{ s}]$ was discretized into 800 elements ($\Delta t = 0.025 \text{ s}$). For the original S2 formulation, we set the

initial concentration to $C_I = 0.001$ (which corresponds to $f_{Hb} = 10^{-15}$) to avoid the singularity at $C = 0$. For the regularized formulation, we use $C_I = 0$, and, in order to test the effect of ε_r on the transient hemolysis estimation, we tried $\varepsilon_r = 0.01$ and $\varepsilon_r = 0.001$. In general, the solution for the formulation with $\varepsilon_r = 0.01$ converges faster than with $\varepsilon_r = 0.001$ by taking 2 or 3 fewer Newton steps per time step.

Figure 17(a) shows the inlet flow rate and the average outlet NIH flux over time. The NIH flux approaches the steady-state value at about $t = 6$ s. The NIH flux shows periodic behavior about one second after the onset of the flow pulsation. Moreover, the lowest NIH value is about the same as the steady-state value indicating that the pulsatile flow leads to higher blood damage.

A single cycle of the NIH flux curve in the periodic state is shown in Fig. 17(b). Note that the results are indistinguishable for different formulations and ε_r values.

Unexpectedly, the peak of the NIH curve appears before that of the flow rate curve. Furthermore, the shape of the NIH curve is distorted; the time difference between the highest peaks $\Delta t = 0.05$ s is shorter than the difference between the lowest peaks $\Delta t = 0.1$ s. Moreover, the mean $\text{NIH}_{\text{avg}} = 1.645 \times 10^{-5}$ g/100L blood over time is higher than the steady-state value, even though the maximum $G_f = 326 \text{ s}^{-1}$ during the transient computation is almost the same as that of steady flow. This is due to traveling vortices in the downstream blood vessel containing high NIH value.

For a single cycle, the NIH contour plot is similar to the streamline plot, as shown in Fig. 18; the regions of high NIH coincide with the regions of flow recirculation. The streamline plot for the highest NIH value, Fig. 18(b), has fewer regions of flow recirculation inside the downstream blood vessel than in Fig. 18(d) for the lowest NIH value. This result strongly suggests that the high outlet NIH flux comes from ejection of the traveling vortices. For example, the damaged blood from the lower surface of the bypass graft flows into Vortex 3 and gets trapped, as shown in Fig. 18(g) and (h). The highly damaged blood can arrive at Vortex 5 through the downstream impact point (D) of Fig. 14. Therefore, Vortex 5 contains the highest NIH value during the cycle, and the maximum NIH flux at the outlet depends on ejection of that vortex.

Furthermore, NIH values inside Area I and II are lower than that of the steady-state result, because of the extremely low shear rate inside these areas. Moreover, the given computation time 20 s is not enough to hemolyze RBC completely. Except for Vortex 2 and 3, which have NIH influxes due to the pulsatile flow, the rest of Area I and II contains virtually no PF hemoglobin during the entire computation.

6. Conclusion

We present a method of solving a convection-reaction system using the space-time least-squares finite element method (ST/LSFEM). To prevent the negative concentration problem, we transform the concentration variable into a squared quantity. Even though the proposed method does not yield a completely diffusion-free solution, it can be easily applied to complex flows while providing enough accuracy for predicting hemolysis in medical devices.

However, there is a room for improvement in the proposed method. For example, eliminating oscillation in the concentration profile can be an important extension of the method. A good candidate is the discontinuity-capturing method developed for the GLS method [27]. Handling discontinuity across the separating streamline can be another area of improvement, especially for the steady-state computation. One may find a way to align mesh

boundaries and use discontinuous basis functions to capture the discontinuity. Another potential candidate is an extended finite element approach based on locally enriching basis function space near the discontinuity, as discussed in Chessa and Belytschko [28].

The proposed method is used for blood damage estimation in two examples: the circular Couette system with a blade and the arterial graft flow. Shear rate and NIH plots provide rich information about the locations of “dangerous” regions, and blood damage distributions inside a given flow. This information can be useful in improving medical devices. Results from the examples we showed warn of the danger of flow recirculations in such devices. Even though some recirculations provide relative low shear rates, they can completely hemolyze RBC given sufficient time, as predicted by analysis of the steady-state hemolysis. In other words, they can trap and hold damaged blood as shown in transient computation. Under pulsatile flow conditions, these recirculations can also travel inside the flow, and can even be split and discharged into the regions of non-recirculating flow. This behavior increases the average NIH flux at the outlet of medical devices.

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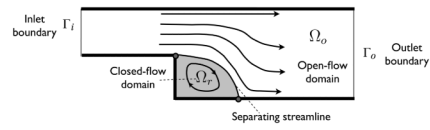


Figure 1. Flow domain consists of an open-flow domain and a closed-flow domain. Note that there is no flux across the separating streamline.

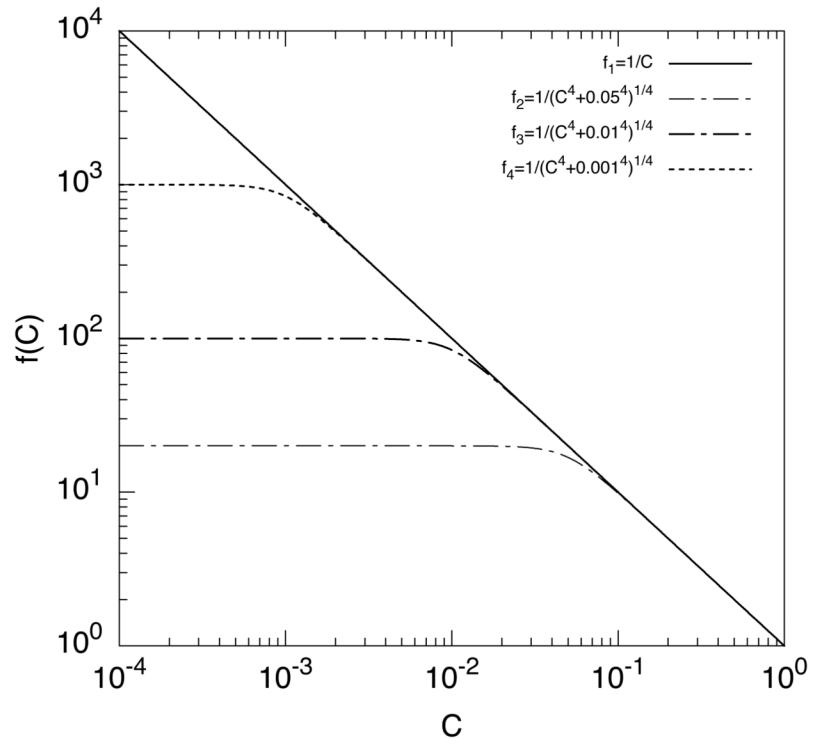


Figure 2. Effect of regularization of $1/C$ with respect to the parameter ϵ_r . f_1, f_2, f_3 and f_4 have $\epsilon_r = 0, 0.05, 0.01$ and 0.001 , respectively.

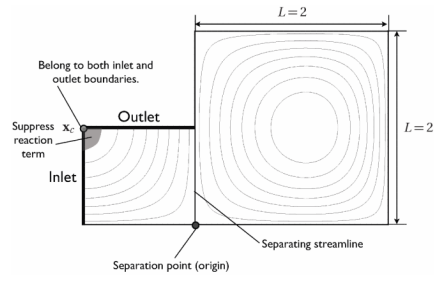


Figure 3.
Recirculating flow streamline plot

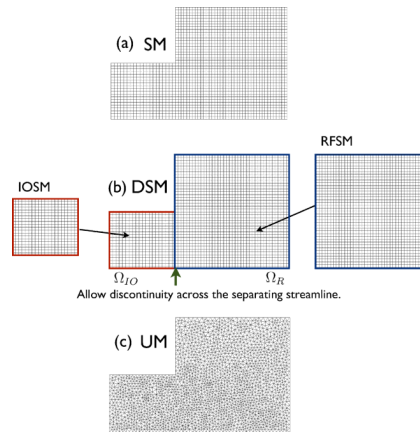


Figure 4. The three different meshes for the recirculating flow. SM, DSM and UM stand for structured mesh, discontinuous structured mesh and unstructured mesh. DSM consists of two separate structured meshes for inlet/outlet flow domain Ω_{IO} IOSM and recirculating flow domain Ω_R RFSM. SM consists of 2000 quadrilateral elements and 4202 nodes. UM consists of 4138 triangular elements and 4298 nodes. DSM is similar to SM, but it allows discontinuity in variable across the separating streamline.

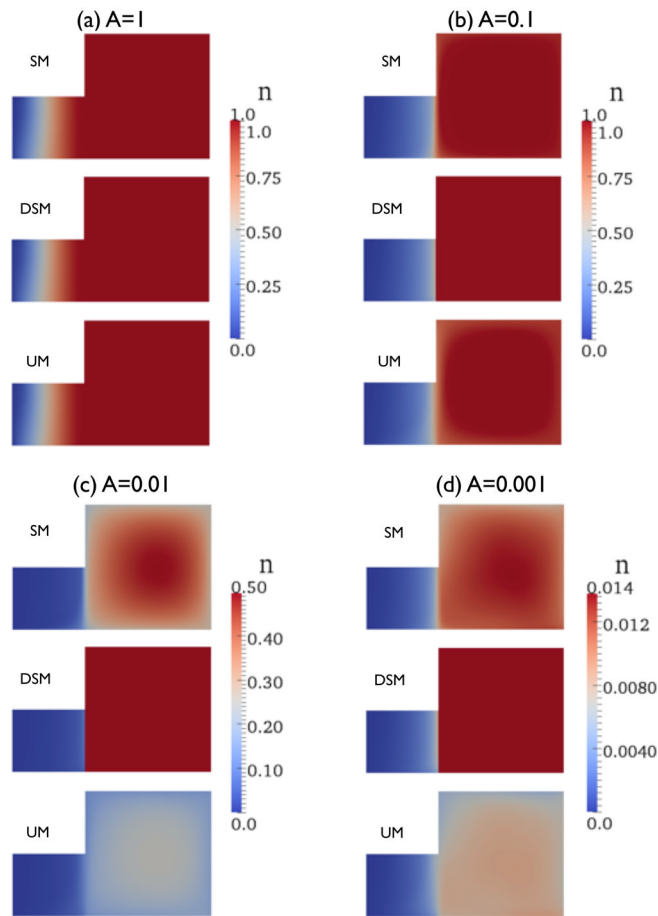


Figure 5. Normalized concentration n plot for different A . Color of contour plot is based on minimum and maximum values of SM for each A and $n = 1$ for Ω_R (RFSM part) of DSM. Note that results from SM and UM for $A = 0.001$ has negative concentration near the upper right part.

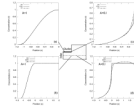


Figure 6.
Normalized concentration n profile for $A = 1$ and $A = 0.1$.

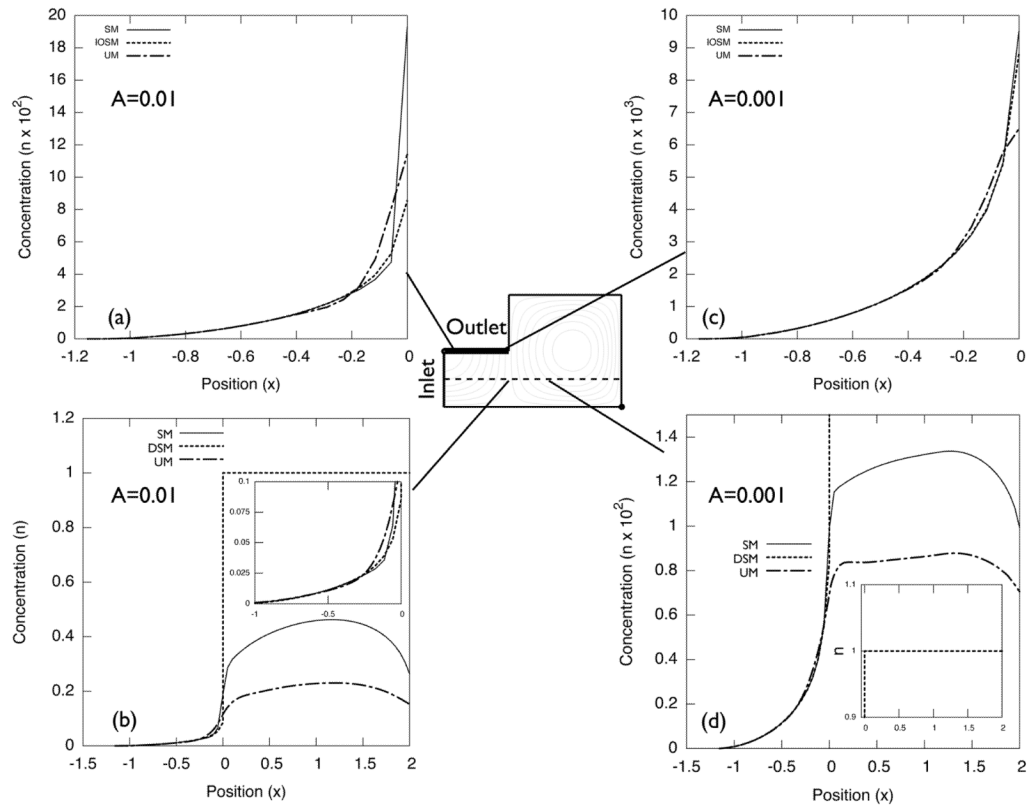


Figure 7. Normalized concentration n profile for $A = 0.01$ and $A = 0.001$.

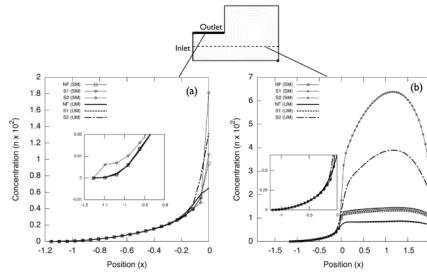


Figure 8. Normalized concentration n profile for different NF, S1, and S2 formulation.

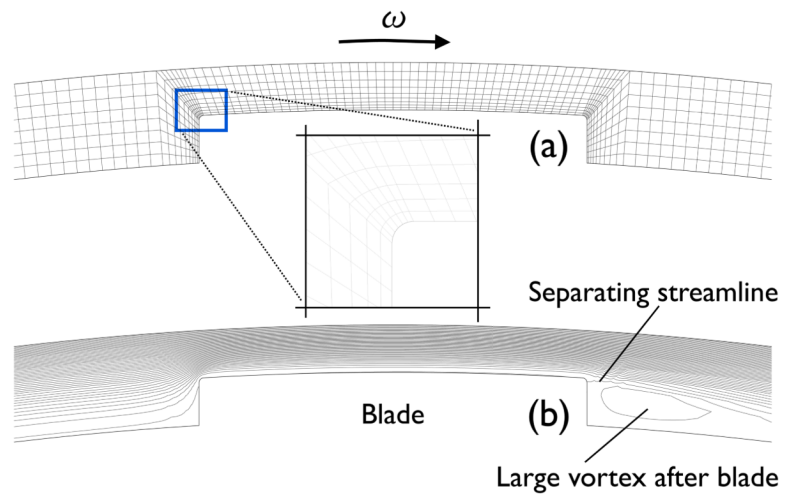


Figure 9. Mesh (a) and streamline plot (b) near the blade of the circular Couette system with blade configuration. Radius of curvature of the rounded corner is $R_r = 50 \mu\text{m}$. 5805 linear quadrilateral finite elements are used to discretize the flow system.

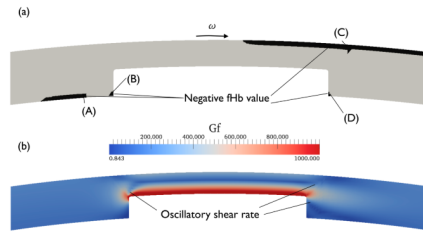


Figure 10. Negative f_{Hb} (a) and projected shear rate G_f (b) plots near the blade of the circular Couette system at $t = 1$ sec from NF formulation at $t = 1$ s. Other than red color regions of (a) represent negative f_{Hb} . In order to contrast oscillatory behaviors of G_f , the maximum value of G_f in (b) is limited to 1000 1/s.

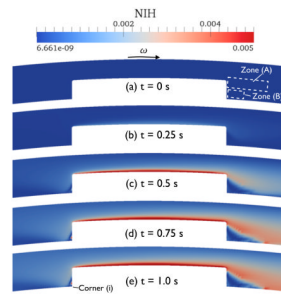


Figure 11. Evolution of NIH value in the circular Couette flow with blade configuration as time progresses.

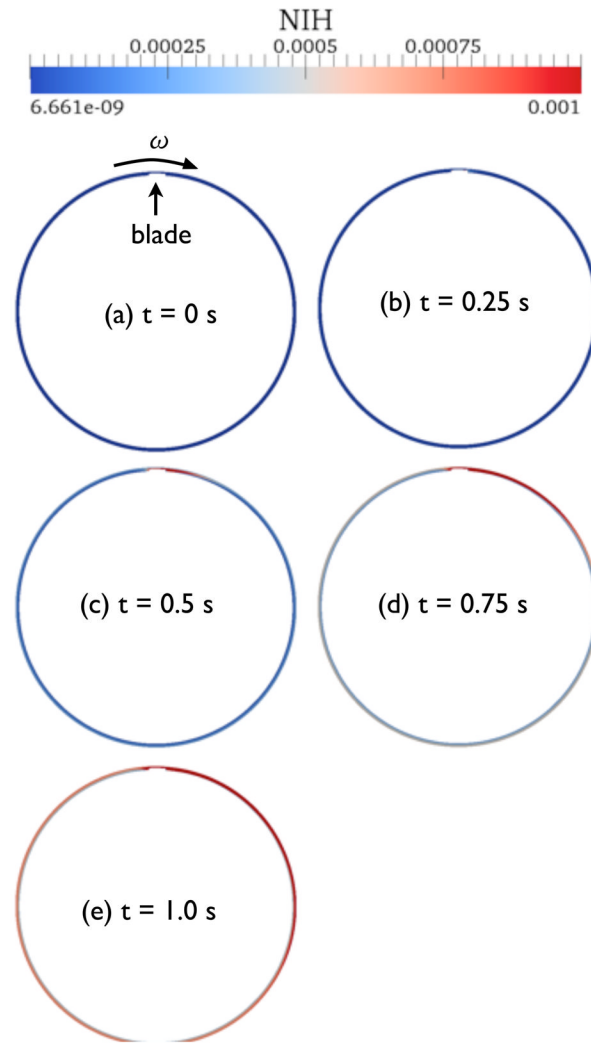


Figure 12. Evolution of NIH value in the circular Couette flow with blade configuration as time progresses.

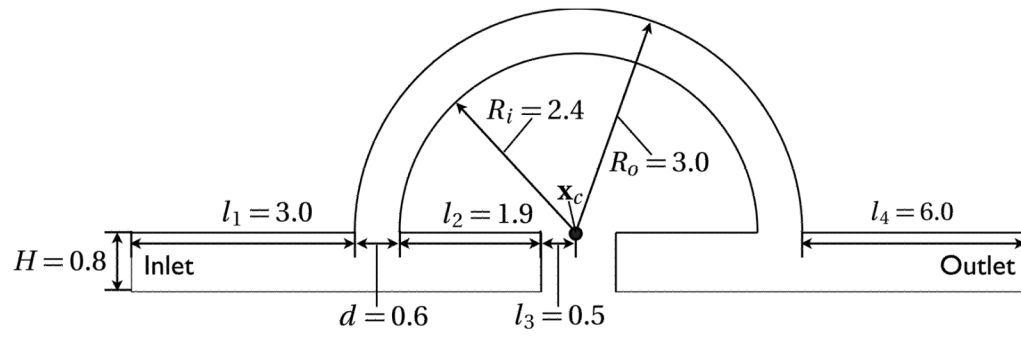


Figure 13.
Geometry parameter for artery graft flow.

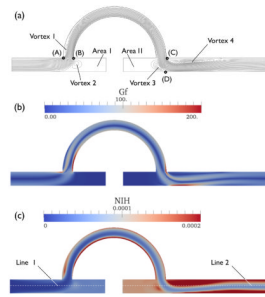


Figure 14.
 Streamline, NIH and G_f contour plots.

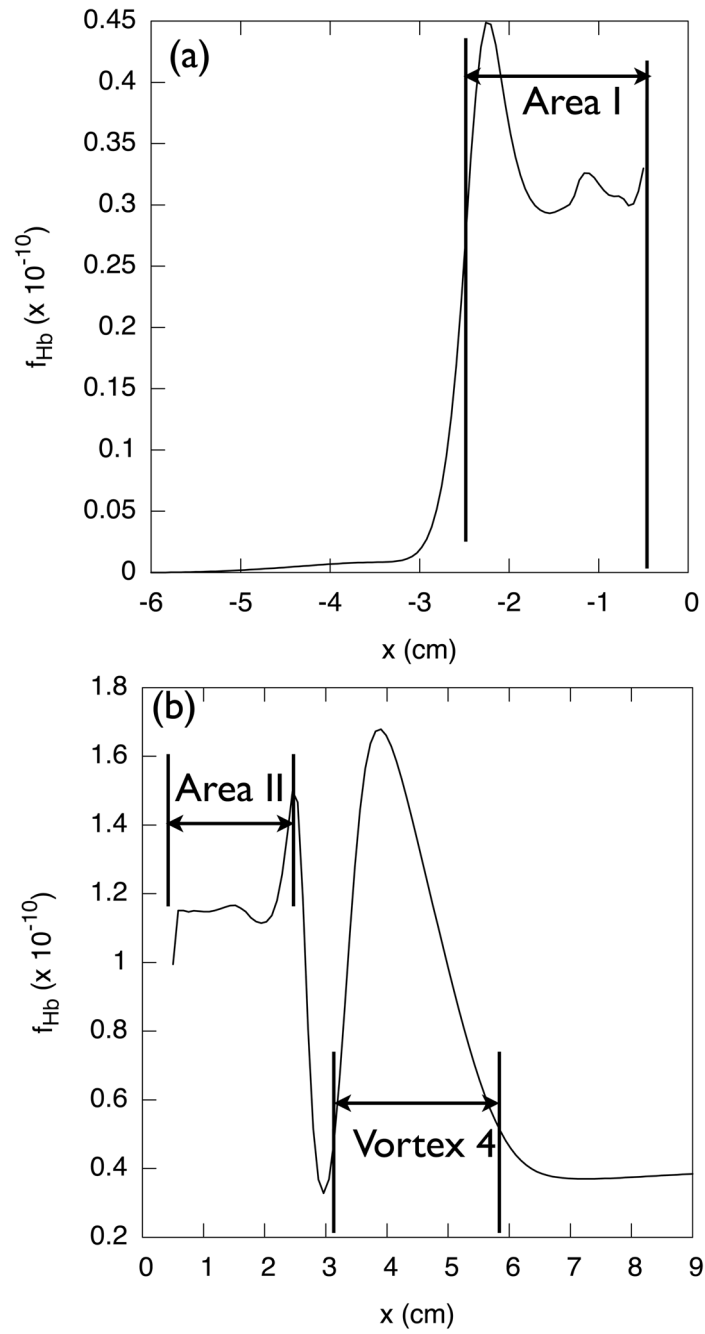


Figure 15. f_{Hb} values along line 1, plot (a), and along line 2, plot (b).

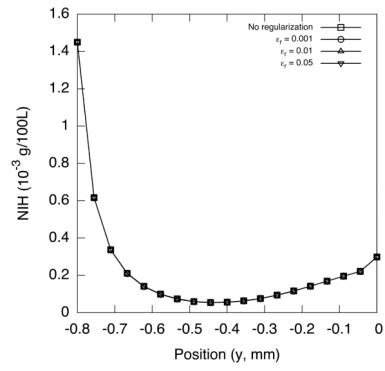


Figure 16.
 NIH value profile at the outlet for different ϵ_r .

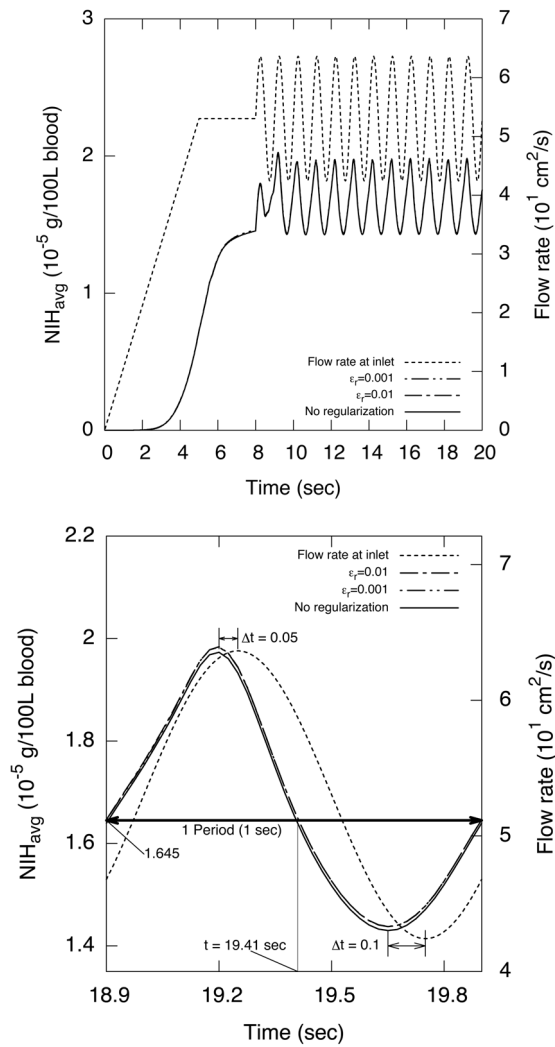


Figure 17. NIH flux across the outlet boundary and pulsatile flow rate at the inlet in the transient artery graft flow for $N_{Re} = 300$. Plot (b) shows one period of NIH flux across the outlet between $t = 18.9$ sec and $t = 19.9$ sec. Note that both NIH flux values were computed using S2 formulation with different regularization parameter values ϵ_r .

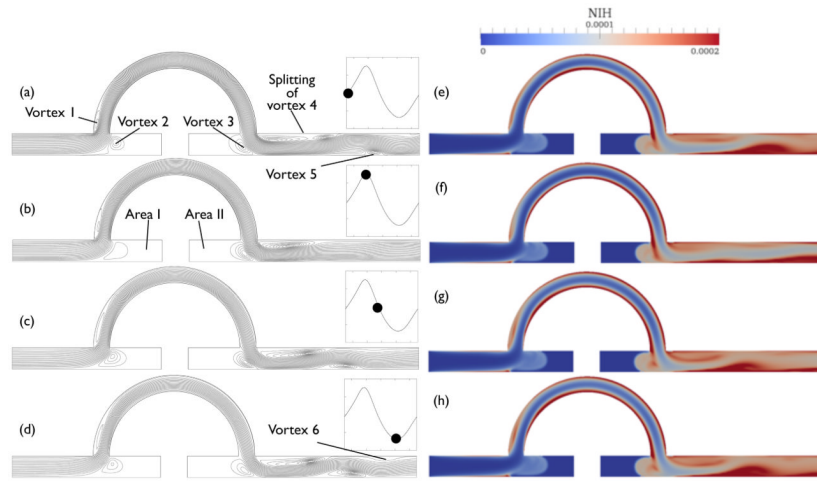


Figure 18. Evolution of the scalar shear rate, from (a) to (d), and the NIH values, from (e) to (h), in one cycle.

Table 1

Raw residuals for least-squares formulation with Eq. (6)

Raw residuals	Expressions
$R_{NF}(f_{Hb})$	$\frac{\partial f_{Hb}}{\partial t} + \mathbf{u} \cdot \nabla f_{Hb} - r_{\Delta Hb}^I (1 - f_{Hb})$
$R_{S1}(C)$	$C \left(\frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C \right) - \frac{\kappa^2 r_{\Delta Hb}^I (1 - f_{Hb}^b - C^2)}{2}$
$R_{S1}(C)$	$\left\{ \begin{array}{l} \frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C - \frac{r_{\Delta Hb}^I}{2} \left(\frac{\kappa^2 (1 - f_{Hb}^b)}{C} - C \right) \text{(S2 formulation)} \\ \frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C - \frac{r_{\Delta Hb}^I}{2} \left(\frac{\kappa^2 (1 - f_{Hb}^b)}{(C^4 + \epsilon_r^4)^{1/4}} - C \right) \text{(regularized S2 formulation)} \end{array} \right.$

Table 2

Sensitivity of raw residuals with respect to f_{Hb}^i or C^i with Eq. (6).

Sensitivity	Expressions
$\frac{\partial R_{NF}(f_{Hb}^h)}{\partial f_{Hb}^i}$	$\frac{\partial \varphi^i}{\partial t} + \mathbf{u} \cdot \nabla \varphi^i + r_{\Delta Hb}^l \varphi^i$
$\frac{\partial R_{S1}(C^h)}{\partial C^i}$	$\varphi^i \left(\frac{\partial C}{\partial t} + \mathbf{u} \cdot \nabla C \right) + C \left(\frac{\partial \varphi^i}{\partial t} + \mathbf{u} \cdot \nabla \varphi^i \right) + r_{\Delta Hb}^l C \varphi^i$
$\frac{\partial R_{S2}(C^h)}{\partial C^i}$	$\left\{ \begin{aligned} & \left(\frac{\partial \varphi^i}{\partial t} + \mathbf{u} \cdot \nabla \varphi^i + \frac{r_{\Delta Hb}^l}{2} \left(\frac{\kappa^2 (1 - f_{Hb}^b)}{C^2} + 1 \right) \right) \varphi^i \text{ (S2 formulation)} \\ & \left(\frac{\partial \varphi^i}{\partial t} + \mathbf{u} \cdot \nabla \varphi^i + \frac{r_{\Delta Hb}^l}{2} \left(\frac{\kappa^2 (1 - f_{Hb}^b) C^3}{(C^4 + \epsilon_r^4)^{5/4}} + 1 \right) \right) \varphi^i \text{ (regularized S2 formulation)} \end{aligned} \right.$