A Multiscale, Biophysical Model of Flow-Induced Red Blood Cell Damage

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A new model for mechanically induced red blood cell damage is presented. Incorporating biophysical insight at multiple length scales, the model couples flow-induced deformation of the cell membrane (\sim 10 µm) to membrane permeabilization and hemoglobin transport (\sim 100 nm). We estimate hemolysis in macroscopic (above \sim 1 mm) 2-D inhomogeneous blood flow by computational fluid dynamics (CFD) and compare results with literature models. Simulations predict the effects of local flow field on RBC damage, due to the combined contribution of membrane permeabilization and hemoglobin transport. The multiscale approach developed here lays a foundation for a predictive tool for the optimization of hydrodynamic and hematologic design of cardiovascular prostheses and blood purification devices. © 2014 American Institute of Chemical Engineers AIChE J, 60: 1509–1516, 2014

Keywords: hemolysis modeling, computational fluid dynamics (CFD), red blood cell, membrane poration

Introduction

In many intra- and extracorporeal biomedical devices, such as hemodialyzers, prosthetic heart valves and ventricular assist devices (VADs), blood is forced to flow through

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artificial ducts, where mechanical stresses can be up to two orders of magnitude higher than those acting in physiological blood flow, 1,2 increasing the risk of red blood cell (RBC) damage. These stresses can cause hemolysis, i.e., the release of hemoglobin (Hb), into plasma. In severe cases, hemolysis causes a decrement in RBC count and it may cause hemolytic anemia. Even in milder, sublytic instances, where Hb is released through temporary pores in the RBC membranes, other pathologies may arise, including reduction of RBC lifetime³ and renal failure.⁴ The design of biomedical devices must account for potential hemolysis to ensure not only efficacy, but also hemocompatibility.⁵ Several experimental studies have been devoted to assessment of RBC damage^{6,7}; they are typically carried out by flowing blood in mockcirculation loops in series with the perfused artificial device and measuring the amount of plasma-free hemoglobin (pfHb) at the end of the experiment. However, such experimental investigations can be slow and expensive because each design iteration requires the construction of a new device, and test results do not provide strong mechanistic links between design alteration and performance changes. Moreover, it is challenging to apply in vitro experimental results directly to in vivo situations.

Therefore, there is a need for a reliable quantitative and mechanistic model to assess device hemocompatibility. Such a model should handle the physics of hemolysis within reasonable computation time, allow for general results (i.e., not confined to specific experimental conditions), and identify potential device drawbacks in the early design phase, enabling both hydrodynamic and hemodynamic computerassisted design of devices.

Early studies on hemolysis⁸ identified shear stress and stress exposure time as the major factors in mechanical hemolysis. A class of "stress-based" models was developed based on the experimental correlation of measured pfHb concentration, shear stress (τ) , and exposure time (t). The most common model is a power-law for the index of hemolysis (IH)

$$IH = \frac{\Delta H b_p}{H b_B} = A t^a \tau^b \tag{1}$$

where $\Delta Hb_p = Hb_p - Hb_p^0$ is the increase in pfHb concentration above the physiological value, Hb_p^0 is the total blood Hb concentration and A, a, b are empirical constants. 9,10

However, this approach has several limitations. The correlations are based on data obtained by exposing blood to steady Couette flow, i.e., uniform and time-constant shear stress in the range 50-255 Pa, within a given exposure time (7–700 ms). Thus, these models are not predictive of general flow conditions: in a typical medical device the shear stress is not a simple scalar, is spatially inhomogeneous and timedependent. In many practical situations, the magnitude of shear stress may exceed the range experimentally measured for the correlations. Finally, this solely empirical approach does not consider fundamental relationships among shear rate, cell shape distortion and hemoglobin release.

In contrast, a "strain-based" approach considers the flowinduced RBC deformation up to the membrane rupture limit, and relates hemolysis to the deformation. In the model proposed by Arora et al.11 RBC deformation is tracked along pathlines to compute the equivalent scalar shear rate experienced by the cell and then the rate of hemolysis based on the extent of cell areal deformation. Chen et al. 12 proposed a damage model based on a threshold strain and compared

Table 1. Properties of RBCs (at 300 mOsm, 37°C)

	Value
Diameter [µm]	8
Cell thickness [µm]	2
Surface area [µm ²]	135
Volume [μm ³]	100
Membrane thickness [nm]	5-10
Shear modulus [mN m ⁻¹]	6.10^{-3}
Bending modulus [Nm]	$2 \cdot 10^{-19}$
Elastic compressibility modulus [mN m ⁻¹]	400

From Guido and Tomaiulo17

model predictions to hemolysis data in hypodermic needles. Even though these two strain-based models 11,12 include more mechanistic information on hemolysis, they lack a description of membrane-level phenomena and do not capture the processes leading from partial damage to irreversible rupture, the so-called sublytic phase of RBC damage.

Arora et al.¹³ defined the following fundamental criteria for formulating models for hemolysis estimation (1) physical properties of RBCs and their variety from species to species, (2) blood flow conditions, (3) RBC behavior in general flows, (4) ease of implementation in complex flow configurations, and (5) ease of integration into computational fluid dynamics (CFD) analysis. Here we propose a sixth important criterion: the model must capture biophysical processes associated with RBC damage. In this work, we propose a novel model for RBC damage estimation in general flows based on a multiscale, biophysical (MSBP) approach, which accounts for both blood fluid dynamics and RBC membrane biophysics.

Background: Mechanical Properties of RBCs and **Membrane Poration**

RBCs at rest are axisymmetric, biconcave, disk-shaped bodies. The main physical and mechanical properties of human RBCs are summarized in Table 1. RBCs are compliant to shearing and bending but remarkably resistant to area stretching (Table 1). Due to the high-stretch resistance, the maximum area strain (a) that RBCs can tolerate ranges from 4%¹⁴ to 6.4%.¹⁵

In flow, RBC membrane deforms in response to fluidinduced viscous forces, developing tension. At low tension, the membrane can deform without area dilation, because RBCs have 40% excess surface area with respect to a sphere enclosing the same volume. 16,17 As tension grows and area strain increases, the formation of membrane pores becomes energetically favorable. Pores relax the membrane tension and quickly reach a stable configuration—in a few µs for pores with radius up to 100 nm. 18 Further deformation and pore nucleation can occur up to the critical area strain for membrane rupture.

Temporary permeabilization of biological membranes under the action of strong flows was shown through the uptake of ~10 nm molecules, 19 and AFM observations of unbroken RBCs after shearing²⁰ showed altered membrane morphology accompanied by an increase in pfHb concentration.

Model Formulation

In our modeling approach flow-induced RBC damage consists of three main processes (Figure 1)

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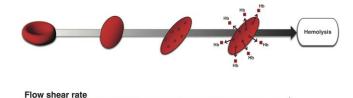


Figure 1. Framework of the MSBP model.

[Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

- 1. Flow-induced RBC deformation and membrane area
- 2. Membrane energy changes and pore formation under deformation.
- 3. Transport of hemoglobin through pores to estimate the index of hemolysis.

Pore formation is assumed to occur instantaneously, because the relative time scale of this evolution $(\approx 10 \text{ µs})^{-1}$ is about four orders of magnitude smaller than the characteristic time scale for RBC deformation ($\approx 100 \text{ ms}$).²¹

The definitions and values of all the constants used in the model are reported in Table 2.

RBC deformation and area dilation in fluid flow

RBCs immersed in a flow exhibit complex motion and deformation patterns. When blood is at rest, cells aggregate in stacks called rouleaux.²² When blood is exposed to a shear flow with low-shear rate (G_f), rouleaux disaggregate and cells tumble as naturally buoyant capsules, while preserving their biconcave shape. As the fluid shear rate increases, tumbling ceases and cells deform into ellipsoidal shapes, with the major axis leaning along the direction of flow.²³ In shear flow this occurs at a shear stress of about 1 Pa. In this regime, the membrane starts to rotate around the cytoplasm, in the so-called tank-treading motion. 24,25 Higher shear rate causes membrane dilation. In steady shear, 8 the critical area strain (6%) is reached when $G_{\rm f}$ approaches 42000 s⁻¹. Below this threshold strain, RBCs can completely recover their biconcave shape after the removal of loading.

The tensor-based model of Arora et al. 11 predicts the morphology of RBCs along their trajectories by considering stretching and contraction, membrane relaxation, rotation and tank-treading. In that work, based on the Maffettone and Minale²⁶ model for droplets, a symmetric, positive-definite tensor S is used to describe cell shape and orientation. The temporal volume-conserving cell deformation is

$$\frac{d\mathbf{S}}{dt} - [\mathbf{\Omega} \cdot \mathbf{S} - \mathbf{S} \cdot \mathbf{\Omega}] = -f_1[\mathbf{S} - g(\mathbf{S})\mathbf{I}] + f_2[\tilde{\mathbf{E}} \cdot \mathbf{S} + \mathbf{S} \cdot \tilde{\mathbf{E}}] + f_3[\tilde{\mathbf{W}} \cdot \mathbf{S} - \mathbf{S} \cdot \tilde{\mathbf{W}}]$$
(2)

where $\tilde{\mathbf{E}}$ and $\tilde{\mathbf{W}}$ are, respectively the rate of strain and vorticity tensors in the frame of reference of the cell rotating with rate Ω with respect to a laboratory fixed frame, and g(S) = 3 III/II, where III and II are the third and second invariants of S. Specifically, Ω is the rate of rotation of the eigenvectors of S (see Ref.11 for details). The parameters $f_1 = 5s^{-1}, f_2 = f_3 = 4.3 \cdot 10^{-4}$, were determined by matching the model with experimental data and incorporate effects of relaxation time, long-lived shape oscillations, tank-treading and critical area strain limit.²⁷ The area of the ellipsoidal cell shape is computed with the method of Keller and

Membrane energy landscape under stretching and quantification of porated area

Flow-induced RBC deformation perturbs the equilibrium configuration of the membrane. Nucleation of pores is dictated by the membrane energy landscape of the perturbation. The configuration of the lipid bilayer is governed by two competing forces: hydrophobic attraction of the lipid tails and headgroup repulsion. The interfacial free energy per molecule is 29,30

$$\mu = \sigma a_{\rm m} + D/a_{\rm m},\tag{3}$$

where σ , $a_{\rm m}$ and D are the interfacial energy at the hydrocarbon-water interface, the interfacial area per molecule exposed to aqueous phase and a constant, respectively. The surface area per molecule of an undeformed membrane $a_0 = (D/\sigma)^{1/2}$, minimizes Eq. 3.

When a nonporated bilayer is stretched, it expands elastically and the total energy of the 2M lipid molecules in the membrane becomes

$$W_{\rm IN} = 2M\mu = 2\sigma A_{\rm m} \left(1 + \frac{A_0^2}{A_{\rm m}^2} \right) = 2\sigma A_0 \left(2 + \frac{\alpha^2}{1 + \alpha} \right),$$
 (4)

where A_0 and $A_{\rm m}$ are the undeformed and total stretched area of the lipid bilayer and $A_{\rm m}$ is the area strain.

The area expansion builds up tension in the membrane, but pores can relieve the tension and minimize the total free energy. Pore formation involves two competing effects^{31,32}

- 1. Decrease in the total energy due to the decrease in the area subjected to the interfacial stress, which tends to expand the pore.
- 2. Increase in the total energy due to the exposure of hydrophobic tails to water along the pore perimeter, which tends to shrink the pore.

Both effects depend on the rigidity $\xi \ge 1$ of the hydrophobic lipid tails.³² Considering the contribution of pores to the total free energy of the membrane yields

$$W_{\rm P} = \gamma L_{\rm P} + 2\sigma' (A_{\rm m} - A_{\rm P}) \left[1 + \frac{A^2_0}{(A_{\rm m} - A_{\rm P})^2} \right],$$
 (5)

where $\gamma = 2h_{\rm ne} \sigma'$ is the pore edge energy, $h_{\rm ne}$ is the nonexposed lipid tail length and $\sigma' < \sigma$ is the apparent hydrophobicity considering the stretching of lipids.³² Pore edge energy is not constant, because h_{ne} varies with α . Following Fournier and Joos, 32 h_{ne} can be estimated by assuming that the hydrophobic surface exposed to water is represented by a cone of lateral surface $S = \xi(a - a_0) = \pi r_L h_e$, where r_L is the radius of lipids, and h_e is the length of the tails exposed to water. The nonexposed length is, therefore $h_{\rm ne} = h_t - \xi r_{\rm L} \alpha$, where h_t is the total length of lipid tails.

We consider *n* circular pores with radius r_p , perimeter L_P = $2\pi nr_{\rm P}$ and area $A_{\rm P} = \pi nr^2_{\rm P}$. In the limit of small porated area, the normalized energy per unit area for a generic pore radius r_0 can be conveniently expressed as a fourth order polynomial³⁰

$$\hat{W}_{N} = \hat{W}_{P} - \hat{W}_{IN} \approx AN^{2}r^{4}_{0} - BNr_{0}^{2} + CNr_{0},$$
 (6)

where $N=n/A_{\rm m}$, $A=2\sigma'\xi-\sigma'_0$, $B=\sigma'_0\pi$, $C=2\pi\gamma$ and

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Table 2. Values of the Model Parameters

Value	Definition and source			
Parameters reported in or estimated from literature sources				
	Relaxation parameter ^{11,27}			
$4.3 \cdot 10^{-4}$	Steady shear deformation parameters ^{11,27}			
$135 \mu m^2$	Unstretched area of RBCs ¹⁷			
$20-50 \text{ mJ m}^{-2}$	Interfacial energy at water- hydrocarbon interface ²⁹			
3.2 mJ m^{-2}	Apparent hydrophobicity ³²			
8	Lipid tail rigidity ³²			
2.5 nm	Length of phospholipid tails ²⁹			
$9.5 \ 10^{-2} \ \text{nm}$	Radius of phospholipid heads ²⁹			
0.16%	Threshold area strain for pore formation ³⁵			
6%	Threshold area strain for complete hemolysis ^{8,15}			
$7.4 \cdot 10^{-3} \ \mu m^{-2}$	Minimum pore density (model parameter)			
$111 \; \mu m^{-2}$	Maximum pore density (model parameter)			
40%	Blood hematocrit			
3.5 mPa s	Blood viscosity			
1056 kg m^{-3}	Blood density			
	s reported in or estin 5 s ⁻¹ 4.3·10 ⁻⁴ 135 µm ² 20–50 mJ m ⁻² 3.2 mJ m ⁻² 8 2.5 nm 9.5 10 ⁻² nm 0.16% 6% 7.4·10 ⁻³ µm ⁻² 111 µm ⁻² 40% 3.5 mPa s			

Parameters obtained by fitting the model to the data of Ref. 37 (Note: these parameters appear in Eq. 15). $h 1.67 \cdot 10^{-8} \mu \text{m s}^{(k-1)}$ Mass transpose Mass transport coefficient prefactor

k 1.54 Power law dependence of mass transport coefficient on shear rate

$$\sigma_0' = \frac{\partial W_{\rm P}}{\partial A_{\rm P}} \bigg|_{A_{\rm P} = 0} = 2\sigma' \xi \left[1 - \frac{1}{\left(1 + \alpha \right)^2} \right] \tag{7}$$

is the surface tension of the intact membrane.30

For $r_0 > 0$, $\hat{W}_N(r_0)$ has a local minimum, corresponding to a stable equilibrium condition. At this minimum, pores with $r_0 = r_p$ form due to membrane deformation.

To close the system of equations, N must be related to the other variables. Eq. 6 requires the condition $N < N_C = 8/27$ $[B^3/(AC^2)]$ for any α to admit the local minimum and it yields equilibrium porated configurations. We assume that Ndepends on strain, because area strain drives the poration. Electroporation studies³³ showed that the equilibrium pore density on membrane pores follows a Boltzmann's type distribution. Because transmembrane voltage and area strain have similar effects on membrane poration,³⁴ we use the same Boltzmann-type dependence of \hat{N} on area strain

$$N = N_0 exp(\zeta \alpha), \tag{8}$$

Equation 8 can be rearranged as

$$N = N_1 exp \left[\frac{\alpha - \alpha_1}{\alpha_1 - \alpha_2} \ln \left(\frac{N_1}{N_2} \right) \right], \tag{9}$$

where N_1 and N_2 are pore densities at area strains α_1 and α_2 , respectively. We set $N_1 = N_c(\alpha_1)$ such that at the critical condition ($\alpha = \alpha_1$), each RBC has exactly one pore. This condition is satisfied when $\alpha_1 = 0.16\%$, which also implies the existence of a minimum threshold strain below which no pores are created. The corresponding shear rate in steady shear is $G_f = 3700 \text{ s}^{-1}$, calculated from Eq. 2. This assumption is supported by experimental observations³⁵ that pfHb does not increase when samples are sheared in the range $3000 - 3500 \text{ s}^{-1}$.

The value of N_2 comes from the membrane rupture condition. At the limit area strain $\alpha_2 = 6\%$ the relative pore area³² is $A_{\rm P}/A_{\rm m} \approx 5.5\%$. At N_2 , the first derivative of Eq. 6 must vanish to satisfy the minimum energy condition. Therefore

$$N_2 = \frac{2Br_P - C}{4Ar_P^3} = 111 \text{ } \mu\text{m}^{-2}.$$
 (10)

Here r_p is calculated from $A_P(\alpha_2) = \pi N_2 r_P^2 A_m(\alpha_2) = \pi A_m$ $(\alpha_2)(2Br_P-C)/4Ar_P$. Combining Eq. 6 and 9 yields the total porated area on the membrane, for any value of α .

Transport of hemoglobin through the pores and expression of the Index of Hemolysis

The opening of pores initiates the release of intracellular content into plasma. Sublytic blood damage occurs when pores are large enough to allow Hb to escape ($r_{Hb} \approx 30$ nm). The Hb flux can be expressed in terms of the concentration difference between the cytoplasm and plasma

$$J_{\rm D} = K_{\rm c}({\rm Hb_{RBC} - Hb_{P}}) = K_{\rm c}({\rm Hb_{B} - Hb_{P}}){\rm Hct}^{-1},$$
 (11)

where Hb_{RBC} is the intracellular Hb concentrations, Hct = 40% is the hematocrit, and K_c is the overall masstransport coefficient.

Assuming homogeneous distribution of RBC size and shape, the mass balance for Hb_p in a batch of sheared blood

$$\frac{\mathrm{dHb_p}}{\mathrm{d}t} = \frac{K_c}{(1 - \mathrm{Hct})} \frac{A_{\mathrm{P}}(\alpha(t))}{V_{\mathrm{RBC}}} \left[\mathrm{Hb_B - Hb_p}(t) \right]$$

$$= R(\alpha(t), K_c) \left[\mathrm{Hb_B - Hb_p}(t) \right].$$
(12)

From Eqs. 1 and 12, the rate of hemolysis can be expressed as

$$\frac{\mathrm{dIH}}{\mathrm{d}t} = \frac{1}{\mathrm{Hb_B}} \frac{\mathrm{dHb_p}}{\mathrm{d}t} = R(\alpha(t), K_{\mathrm{c}}) \left(1 - \mathrm{IH} - \frac{\mathrm{Hb_p^0}}{\mathrm{Hb_B}} \right). \tag{13}$$

The term $R(\alpha(t), K_c)$ includes two contributions to the transport across RBC membrane (Figure 2): mass-transport rate, controlled by the overall mass-transfer coefficient K_c , and size of pores, dependent on area strain α .

For an initially undamaged blood sample, Eq. 13 can be integrated with the initial condition IH(t = 0) = 0, to obtain

$$IH(t) \approx 1 - exp \left[-\int_{0}^{t} R(\alpha(t), K_{c}) dt \right].$$
 (14)

In physiological conditions $Hb_p^0/Hb_B \approx 10^{-4} \ll 1$; thus, this term is neglected in Eq. 14.

To estimate the mass-transfer coefficient K_c , the RBC is approximated as small, force-free particle in a shear dominated flow; hence, the rate of mass transport can be expressed as a function of the scalar fluid shear rate³⁶

$$K_{\rm c} = hG^k_{\rm f} \tag{15}$$

Constants h and k were derived from fitting experimental

Results and Discussion

Figure 3 shows the results from the fitting of Eq. 15 against literature data of hemolysis in steady Couette flow.³⁷

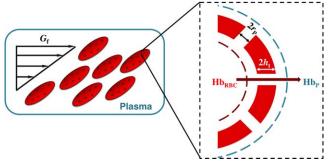


Figure 2. Model of hemoglobin transport in a control volume of blood subjected to fluid shear rate $G_{\rm f}$.

Enlarged is the cross-sectional view of the RBC membrane of thickness containing circular pores of radius. A flux of hemoglobin is established, which is proportional to the overall mass-transport coefficient, and the concentration difference between the cytoplasm and the extracellular plasma (Hb_{RBC}-Hb_P). [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

The original data had to be corrected for two reasons. First, in the data, the hemolysis did not vanish in the absence of shearing, likely due to factors such as contact with exogenous surfaces.³⁸ Second the original definition of shear rate and exposure time did not account for the Poiseuille flow through the device. The analysis and correction procedure is reported in Supplementary Material.

Equation 15 was fit to experimental data at exposure time t=887 ms, which is almost 10 times higher than the characteristic time of RBC deformation dictated by Eq. 2: $t_{\rm d} = (f_2 G_{\rm f})^{-1}$. Therefore, at this exposure time range, the mass-transport rate can be assumed to depend only on the flow field around the cell as expressed by Eq. 15. The optimal fitted values of the constants are $h=1.67 \cdot 10^{-8} \mu {\rm m \ s}^{(k-1)}$, $k=1.54 \ ({\rm RMSE}=6.8 \cdot 10^{-3})$.

Figure 4 shows the comparison between other data sets from the same work 37 and predictions from MSBP model, at varying shear rate and for different values of exposure time. A satisfactory agreement was found for data sets in which $t \gg t_{\rm d}$, whereas model predictions deviate from data when $t \approx t_{\rm d}$. We suspect that such discrepancy might stem from an underestimation the membrane area strain at short exposure times. The parameters in the model of cell deformation (Eq. 2) were determined by considering cell rupture conditions, membrane relaxation time and restricting the parameter range so as to avoid cell shape oscillations in steady shear. This choice allows the model to predict RBC shape for higher shear rates and exposure times typical of cardiovascular devices, such as blood pumps and heart valves $(0.5 < t < 1 \text{ s}, G_{\rm f} > 20,000 \text{ s}^{-1})$.

To illustrate model predictions in complex flows and the integration with large-scale CFD computations, we chose a blood flow system with a curved pipe grafted to a straight channel. The Newtonian fluid velocity field was computed by the Galerkin/least-squares (GLS) finite element method with bilinear basis functions for both velocity and pressure, ³⁹ using the third-order in space, second-order in time, in-house code XNS. Mesh and time-step size were chosen accordingly to a previously established methodology (details on flow governing equations, GLS formulation and flow solution strategy can be found authors' in previous works ^{40–42}). The choice of a Newtonian model for blood is commonly

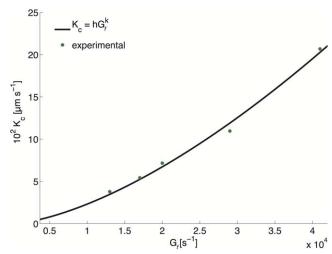


Figure 3. Overall mass-transport coefficient vs. shear rate.

Points represent the values calculated from experimental rate of hemolysis; the curve is the result of fitting hemolysis data from Heuser and Opitz³⁷ at t = 887 ms with Eq. 15. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

adopted when simulating high-shear flows (>100 s⁻¹). In such flow conditions that are typical of extracorporeal devices, it has been shown that the viscosity of human blood at 45% hematocrit and temperature of 37°C reaches a constant value of 3.5–4 mPa s. At lower shear rate, blood exhibits shear-thinning, viscoelastic and thixotropic properties, due to the shear-controlled aggregation and disaggregation of rouleaux and the Newtonian assumption cannot be considered valid. Extensive review of alternative constitutive equations for the description of properties at low shear rates can be found in Ref. 44. Blood flow in the domain Ω with boundary Γ was computed by solving the momentum and mass conservation equations for an incompressible fluid

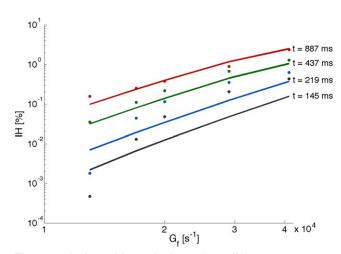


Figure 4. Index of hemolysis at four different exposure times.

Comparison between original Giersiepen data¹⁰ (points) and model results (lines). Data set at t=887 ms was used to fit $K_{\rm c}$, whereas the other three data sets at 145 < t < 437 ms were predicted with the model. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

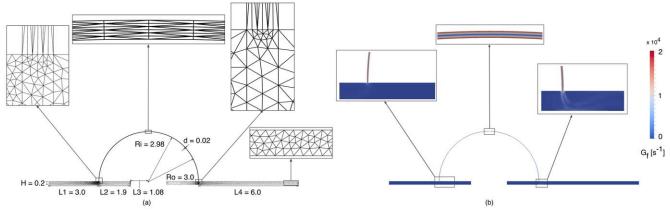


Figure 5. Flow in a curved pipe grafted to a straight channel (a) geometry (units in cm) and mesh, and (b) scalar shear rate profile.

[Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]

$$\rho \left(\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} \right) - \nabla \cdot \mathbf{T} = \mathbf{0}$$
 (16)

$$\nabla \cdot \mathbf{u} = 0 \tag{17}$$

where $\rho = 1056 \text{ kg m}^{-3}$ and $\mu = 3.5 \text{ mPa}$ s are blood density and viscosity, **u** is the velocity, $\mathbf{T} = p\mathbf{I} + \mu(\nabla \mathbf{u} + \nabla \mathbf{u}^T)$ is the total stress tensor, and p is the pressure. The computed velocity and pressure were used in Eq. 2 to calculate RBC deformation, 13 and the rate of hemolysis along 50 path lines equally spaced on the inlet section, starting at a distance of 150 µm from the upper wall boundary. Figure 5a shows the twodimensional (2-D) flow domain. The inlet and outlet ports are H = 0.2 cm wide; the curved pipe radius is 200 μ m. The Reynolds number is $Re = \rho U_{max}H/\mu$, where U_{max} is the maximum xvelocity at the inlet boundary. The fluid dynamic problem was solved at Re = 100 by imposing no-slip ($\mathbf{u} = \mathbf{0}$) at the walls, and a fully developed parabolic velocity profile at the inlet $(\mathbf{u}(y) = \mathbf{i}U(y))$ and outlet $(\mathbf{n} \cdot \nabla \mathbf{u} = \mathbf{0})$, where **n** is the outward normal vector). The flow domain was discretized with an unstructured mesh of 8,124 triangular elements, refined near the inlet and outlet of the curved graft to trace particle path

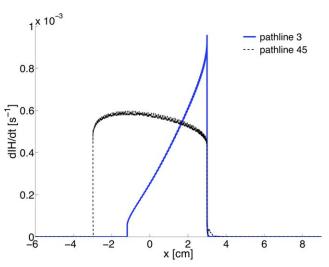
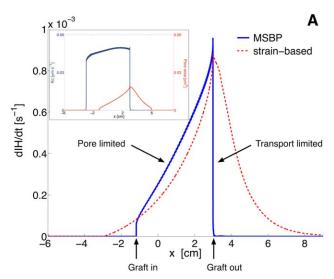


Figure 6. Rate of hemolysis computed with MSBP model on pathlines 3 and 45.

Inlet of the graft is at x = -3, outlet is at x = 3. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary.com.]



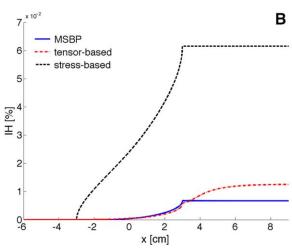


Figure 7. Comparison of results predicted by the different models on pathline 3 (A) rate of hemolysis. Inset: comparison between the overall mass-transport coefficient and pore area on pathline 3.

Inlet of the graft is at x = -3, outlet is at x = 3, and (B) index of hemolysis. [Color figure can be viewed in the online issue, which is available at wileyonlinelibrary. com.]

Table 3. Flow-Rate Averaged Index of Hemolysis, $IH_{avg}\ [\,\%\,]$

	Stress-based	Strain-based	MSBP
Coefficients from eq. S10 in Supplementary	$3.340 \cdot 10^{-2}$	$1.623 \cdot 10^{-2}$	$6.532 \cdot 10^{-3}$
Material Coefficients from Giersiepen et al. ¹⁰	$7.829 \cdot 10^{-2}$	$3.659 \cdot 10^{-2}$	

accurately (Figure 5a). Minor oscillations are caused by the interpolation of velocity \mathbf{u}_n with piecewise linear functions. Figure 5b shows the fluid shear rate profile with a magnification of the zones upstream, downstream and center of the curved graft.

The advantages of the biophysical approach of the MSBP model can observed when the results for two particles initially subjected to same fluid shear rate are compared. Figure 6 shows the rate of hemolysis computed for the particles starting at a distance of 200 µm from the upper (pathline 3) and lower boundary (pathline 45), respectively. For the particle on pathline 3, no hemolysis occurs until the axial location x = -1.2, where the area strain reaches the threshold value and pores open on the membrane. Then the rate of hemolysis steeply increases inside the graft, following the RBC deformation and poration dynamics. On the other end, the particle on pathline 45 deforms almost instantaneously beyond the limit threshold for poration, due the shear pulse near the graft inlet. After this initial surge, the pore area and the rate of hemolysis remain almost constant inside the graft. At the exit of the curved pipe, the rate of hemolysis computed on both pathlines drops suddenly to $\approx 10^{-5} \text{ s}^{-1}$: this effect is due to the decrease of the rate of transport, which is controlled by the fluid shear rate G_f (Eq. 15). Thus, with this novel biophysical description it is possible to separate the factors controlling the hemolysis process: the pore opening mechanism provides an on-off switch for Hb flux and appears to be the rate-limiting step in the high-shear zones of flow domain. The convective mass-transport mechanism controls the intensity of the Hb flux at low shear rate even when RBCs are still relaxing from the deformed state (Figure 7a, inset). Such a description of the interplay between the flow dynamics conditions, the viscoelastic RBC deformation and permeabilization of the cell membrane and the transport of hemoglobin at the sublytic stage of hemolytic damage cannot be captured with current existing models. In Figure 7a the rate of hemolysis on pathline 3 computed with MSBP and tensor-based models are compared. In the tensor-based model the rate of hemolysis depends only on RBC deformation in all of the regions of the flow domain; the MSBP model allows to estimate the relative contribution of the causes of rate change. The total index of hemolysis along pathline 3 calculated with the stress-based, tensor-based and MSBP model is presented in Figure 7b. Stress-based¹⁰ and tensor-based¹¹ models predict nonzero rate of hemolysis in the entire channel, even in the zones where the shear rate is well below the hemolytic thresholds⁴⁵ and, rather, close to values typical of physiological circulation¹ $(\sim 100-200 \text{ s}^{-1})$. Therefore, current models can lead to an overestimation of the index of hemolysis, particularly when stress-based models are used, due to the quadratic, instantaneous dependence of damage on the scalar shear stress.

The total average hemolysis index in the channel was computed using the flow-rate average of IH at the outlet of the graft ($\mathrm{IH}_{\mathrm{avg}}$)

$$IH_{avg} = \frac{\int_{\Gamma_0} IH(\mathbf{u} \cdot \mathbf{n}_0) d\Gamma}{\int_{\Gamma_0} (\mathbf{u} \cdot \mathbf{n}_0) d\Gamma}$$
(18)

Equation 18 was used to compare blood damage predictions from MSBP model, stress-based model and tensor-based model (Table 3). Predictions of IH_{avg} with stress-based approach are 5–10 times higher than the other models, in agreement with findings of other studies.

Conclusion

In this work, we describe a novel approach for estimation of hemolysis in complex blood flows, based on a multiscale, biophysical model. The model consists of three components: deformation and area strain of the cell, formation of membrane pores and transport of hemoglobin. The test of hemolysis prediction in model geometry shows that, through the inclusion on biophysical phenomena, the MSBP model enables the estimation of the multiple flow-induced processes that contribute to cell damage. The model can be integrated with CFD analysis for computational design and optimization of biomedical devices to potentially identify zones at higher hemolytic impact within the flow domain. Its predictive ability is presently limited by the lack of reliable and consistent in-vitro hemolysis measurements³⁸ taken at the short exposure times (< 500 ms) typical of VADs; such experimental data would permit the validation and tuning of the model in fast flows.

Each component of the MSBP model can be refined and improved independently. Ongoing work is directed toward the refinement of RBC deformation model, to include the membrane viscoelastic response to short time shearing and toward enhancing the mass-transfer model.

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Conflict of Interest Statement

The authors declare no conflict of interest associated with this work.

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