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Title: Impact of Incremental Perfusion Loss on Oxygen Transport in a Capillary Network

Mathematical Model

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ABSTRACT

Objectives: To quantify how incremental capillary perfusion loss, such as that seen in experimental models of sepsis, affects tissue oxygenation using a computation model of oxygen transport.

Methods: A computational model was applied to capillary networks with dimensions 84x168x342 (NI) and 70x157x268 (NII) μm , reconstructed *in vivo* from rat skeletal muscle. Functional capillary density (FCD) loss was applied incrementally up to ~40% and combined with high tissue oxygen consumption to simulate severe sepsis.

Results: A loss of ~40% FCD loss decreased median tissue PO_2 to 22.9 and 20.1 mmHg in NI and NII compared to 28.1 and 27.5 mmHg under resting conditions. Increasing red blood cell supply rate (SR) to baseline levels returned tissue PO_2 to within 5% of baseline. High consumption combined with a 40% FCD loss, resulted in tissue anoxia in both network volumes and median tissue PO_2 of 11.5 and 8.9 mmHg in NI and NII respectively; median tissue PO_2 was recovered to baseline levels by increasing total SR 3 – 4 fold.

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Conclusions: These results suggest a substantial increase in total SR is required in order to compensate for impaired oxygen delivery as a result of loss of capillary perfusion and increased oxygen consumption during sepsis.

KEYWORDS

sepsis; oxygen delivery; functional capillary density; blood flow; oxygen transport

ABBREVIATIONS USED

3D, three-dimensional; EDL, extensor digitorum longus; FCD, functional capillary density; NI, network I; NII, network II; PO₂, partial pressure of oxygen; RBC, red blood cell; SO₂, oxygen saturation; SR, supply rate;

INTRODUCTION

Mortality from sepsis and septic shock remains high despite improved diagnosis using biomarkers [21] and implementation of goal directed therapy [22,24,25]. Sepsis has been defined as a systemic inflammatory response to a bacterial infection [4] and is a common cause of mortality in a range of patient groups. Observable changes to blood flow within the microcirculation have been identified clinically within hours of infection onset [29].

Dysfunctional microvascular blood flow has been identified in animal models as a characteristic associated with the systemic inflammation in experimental sepsis; the most

notable aspect of blood flow disruption in these models is the loss of capillary perfusion in tissue remote to the site of infection [2,20]. The loss of perfused capillaries in the microvasculature impairs oxygen delivery and is followed by deleterious effects caused by hypoxia in living tissue [6]. Inadequate oxygen delivery in sepsis has therefore been hypothesized as a factor contributing to multiorgan dysfunction and patient mortality and as such has received considerable attention from researchers and clinicians.

Experimental models have shown varying degrees of perfusion loss in skeletal muscle, liver, brain and intestinal microcirculation [1,6,27,28]. The challenge of making direct measurements of tissue oxygen *in vivo* has led to the use of computer models in order to quantify how loss of capillary perfusion impacts oxygen delivery to living tissue. Previous modeling efforts by Goldman et al. utilized generated parallel capillary arrays and experimental measurements to provide a detailed description of oxygen tension within three-dimensional tissue volumes with loss of functional capillary density consistent with that observed in the skeletal muscle of a rat peritonitis model [12,13]. By adjusting tissue consumption in the model Goldman et al. matched observed oxygen extraction ratios from skeletal muscle capillaries. Goldman et al. concluded that increased consumption as predicted by the model and decreased oxygen supply would make tissue more susceptible to hypoxia in sepsis [13].

In a previous study we demonstrated that computer models using equivalent parallel capillary arrays result in higher estimates of tissue PO_2 when compared to capillary geometries reconstructed from intravital video recordings of blood flow *in vivo* [10]. The discrepancy between calculated tissue PO_2 in parallel arrays and reconstructed network geometries was shown under normal conditions as well as under simulated exercise,

ischemia and hypoxia. We concluded that real network geometries reconstructed from experimental observations, produced oxygen transport solutions that more accurately represent the spatial PO₂ distribution and mean PO₂ values *in vivo* compared to equivalent parallel arrays generated using average measurements from the same experimental data sets. Given this result, computational models should utilize reconstructed or realistic network geometries, and corresponding experimental measurements of velocity, hematocrit and oxygen (SO₂ or PO₂) whenever possible.

In the present study we examine the impact of progressive functional capillary perfusion loss on oxygen transport in 3D reconstructed capillary networks. The impact of incremental FCD loss on oxygen delivery *in vivo* is not known, though computer models utilizing parallel arrays have shown decreased tissue PO₂ as a result of FCD loss [12,13]. By utilizing reconstructed networks we are able to simulate septic flow dysfunction in real capillary networks. Application of a physical flow model was used to determine blood flow redistribution resulting from stoppages in individual capillary segments. We then compared the resulting oxygen transport changes to solutions produced using baseline red blood cell supply rate produced from direct vessel-by-vessel hemodynamic and oxygen measurements made from the sample networks *in vivo*.

Recent clinical studies have monitored venous O₂ saturation in septic patients and used this metric as a target measurement for early goal directed therapies. Central venous SO₂ has been used as a trigger for blood transfusion and has been shown to improve through fluid resuscitation [22,24,26]. In the current study we were able to quantify how increasing red blood cell supply rate affects venular outflow saturation (SvO₂) from discrete

capillary networks. The relationship between SvO₂ and restoration of basal tissue PO₂ levels can be related to clinical measurements of mixed venous saturation as a target for treatment of septic patients.

The objective of the present study is to quantify how varying degrees of FCD loss affect tissue oxygenation. Furthermore we aim to determine if increasing blood flow to hypoxic tissue will ameliorate oxygen delivery in the presence of increased oxygen consumption, a condition observed in experimental models of sepsis. Understanding how oxygen delivery is impaired in this model and how increased blood supply affects tissue oxygenation provides support to the concept of early goal directed therapy.

MATERIALS AND METHODS

3D Network Reconstruction

Two capillary networks from intravital microscopy of rat extensor digitorum longus muscle [20] that were reconstructed for a previous study [9], were used in blood flow and oxygen transport simulations; the methods of which have been described previously [9,30]. Briefly, video sequences of microvascular flow were processed to create functional images and overlapping fields of view and focal planes were registered to produce a mosaic map of the volume of interest. Characteristic features in each image were used for registration allowing for vessels to be tracked between images and focal planes. Vessels in each image were segmented using both manual and automated methods. Connections between vessel segments and at bifurcations were created using automated algorithms creating contiguous maps of network geometries. Completed geometries were processed with quality control software to ensure contiguous geometry, and appropriate proper flow directions as

observed *in vivo*. Resulting vascular maps provide an accurate representation of the structure and transport conditions observed experimentally (as detailed below) in two discrete network volumes (see Table 1 for a summary of the network geometric data).

Capillary Hemodynamic and Oxygen Saturation Measurements

Individual vessels within each network geometry were analyzed to measure hematocrit, cell velocity and oxygen saturation using custom software package similar to that described previously [3,11,18]. Vessel segments were selected from functional images of microvascular flow [17] and space-time images [8] were created from 60 second long video sequences. Space-time images were used to measure cell velocity, hematocrit, and oxygen saturation using dual wavelength absorption spectroscopy [7,19]. Mean values measured for each vessel were indexed to the network geometry such that recorded data measured *in vivo* was paired with the specific capillary within the 3D geometry.

Modeling Flow in 3D Capillary Networks

Network reconstructions were processed to create a node-to-node description of the geometry for use in a steady state flow model [9]. Velocity and hematocrit values not measured experimentally were calculated using a mass balance to determine supply rate in parent and daughter vessels; this process was repeated in upstream segments as necessary to determine inflow node velocity, hematocrit, and supply rate. An existing steady state flow model [14] was applied using the flow equations described by Pries et al [23]. Inlet and outlet pressures were adjusted to approximate the specific velocity distribution

observed in vivo and to create a starting point for further scaling of the flow model. RBC supply rates were recorded from the in vivo data in a cross section of the network and the total network RBC supply rate was determined. The pressure boundary conditions were incrementally adjusted until the RBC supply rate matched the experimentally measured values within $< 0.1\%$ [9].

In order to simulate the loss of functional capillary density (FCD) observed in the experimental models of sepsis, a subset of vessels were set to be stopped for a given perfusion loss using the following approach. Individual stop flow capillaries were randomly selected from the group of vessels within the network that intersected in projection with two sampling lines perpendicular to the network volume Y axis (Figure 1). Blood flow was arrested in the target vessels by increasing intra-capillary resistance such that the resulting flow solution calculated supply rate to be $< 10^{-4}$ of the original values in each of the target stop flow capillaries. Percent of FCD loss was calculated by counting the number of stopped flow vessels divided by the total number of vessels that intersected each sample line. Levels of FCD loss were created to approximately correspond to 10%, 20%, 30% and 40% FCD loss. The original pressure drop across the networks in the incremental perfusion loss simulations were held constant resulting in decreased supply rate with increasing FCD loss (Table 2 and 3). Longitudinal capillary point density for all levels of FCD loss, calculated as described previously[10], are shown in Figure 2. Flow solutions were created as described previously for each network at four levels of FCD loss; the $\sim 40\%$ range of FCD loss at 1X, 2X, 3.5X and 4X baseline RBC supply rates; and control simulations by iteratively adjusting inflow pressures until the target supply rate in inflow vessels was reached [9]. A single counter current inlet vessel in Network II was held constant at the originally measured RBC

supply rate for all network II simulations. Control conditions with 0% FCD loss and supply rate matched to the incremental perfusion loss cases, and high consumption cases at 1X, 2X, 3.5X, and 4X supply rate were generated and separate simulations were run as a comparison for the cases with FCD loss.

Oxygen Transport Model

Oxygen tension in tissue was calculated using a finite difference model of oxygen transport developed by Goldman and Popel [14] and described previously [9]. Flow solutions generated for each case described above were used as the input for the transport models in order to describe tissue PO_2 within the sample volumes under each simulated condition. Finite elements were defined for the entire volume resulting in a monotonic $2 \mu\text{m}$ grid spacing. The diffusion between individual volume elements was calculated at each time step yielding PO_2 for each volume element, $P(x,y,z,t)$, as follows: [9,14].

$$\frac{\partial P}{\partial t} = \left[1 + \frac{c_{Mb}}{\alpha} \frac{dS_{Mb}}{dP} \right]^{-1} \left\{ D \nabla^2 P - \frac{1}{\alpha} M(P) + \frac{1}{\alpha} D_{Mb} c_{Mb} \nabla \cdot \left(\frac{dS_{Mb}}{dP} \nabla P \right) \right\} \quad (1)$$

where D , α , and $M(P)$ are the diffusion coefficient, solubility and consumption rate of O_2 respectively of the tissue. Myoglobin concentration (c_{Mb}), diffusivity (D_{Mb}), and saturation (S_{Mb}) are also addressed through Equation 1 where S_{Mb} is determined by $S_{Mb}(P) = P / (P + P_{50,Mb})$ where $S_{Mb}(P)$ is the myoglobin saturation at a given partial pressure and $P_{50,Mb}$ is the partial pressure at which myoglobin is 50% saturated. Oxygen levels in the blood were determined within each vessel at each axial location (ξ) using a convective mass balance equation that describes blood oxygen saturation $S(\xi,t)$:

$$\frac{\partial S}{\partial t} = -u \frac{\partial S}{\partial \xi} - \frac{1}{\pi R} \left[C + \alpha_b \frac{dP_b}{dS} \right]^{-1} \oint j \cdot d\theta \quad (2)$$

where u is the mean blood velocity, R is capillary radius, j is the oxygen flux out of the capillary at the axial location (ξ, θ) , C is the O₂-binding capacity of blood, P_b is the intracapillary PO₂ and α_b is the solubility of O₂ in plasma.

The flux of O₂ between capillaries and tissue is was defined as:

$$j = \kappa (P_b - P_w) \quad (3)$$

where κ is the mass transfer coefficient and P_w is the tissue PO₂ at the capillary surface. κ is a function of the capillary hematocrit in a given vessel and reflects the effect of red blood cell spacing on diffusional exchange between capillary and tissue [5]. The boundary condition at the capillary-tissue interface was specified as:

$$-D\alpha \frac{\partial P_w}{\partial n} = j \quad (4)$$

where n is the unit vector normal to the capillary surface and j is defined by Eq. 3. Boundary conditions at the tissue edges were specified as a zero flux boundary condition.

The above O₂ transport equations 1 – 4 were combined with Michaelis-Menten consumption kinetics, $M = M_0 P / (P + P_{cr})$ and the Hill equation for oxyhemoglobin saturation, $S(P) = P^n / (P^n + P_{50}^n)$ to define O₂ transport within the 3D volume [15]. The baseline oxygen consumption rate, M_0 , (Table 2 & 3) was selected such that the resulting capillary SO₂ throughout the network fit approximately with experimental observations. Simulations were ran from a starting state of zero PO₂ throughout the volume and diffusion was calculated between capillaries and tissue elements incorporating consumption for each

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case as described in Table 2 and 3. Values for constants used in the equations above are listed in Table 4. Each simulation was run to a steady state condition established when the slope of tissue PO_2 over time in corner elements was equal to zero.

RESULTS

Steady state oxygen transport simulations were completed, for each reconstructed network geometry, under four incremental levels of functional capillary density loss with normal oxygen consumption conditions. Tissue PO_2 distributions calculated for each case under normal consumption are shown in Figure 3. Increasing FCD loss and a concomitant decrease in RBC supply rate resulted in a progressive reduction in median tissue PO_2 levels from baseline values of 28.1 mmHg and 27.5 mmHg for networks I and II respectively. At the highest level of FCD loss (40% for network I and 42.9% for network II) median tissue PO_2 was reduced to 22.9 mmHg, in network I, and 20.1 mmHg in network II. In both network geometries the highest level of FCD loss caused the span of the fourth quartile of PO_2 values to widen and tissue PO_2 values residing in the third quartile to fall below the median value at baseline (Figure 3). When boundary pressures were increased to restore RBC supply rate to baseline levels for 40.0% and 42.9% FCD loss, median tissue PO_2 increased to 26.7 mmHg in network I and 26.8 mmHg in network II; returning median PO_2 to within 5% of baseline levels.

Color PO_2 surface maps in Figure 5 illustrate PO_2 conditions with normal consumption in the network I sample volume under baseline and the resulting drop in PO_2 following a 40% FCD loss denoted in vessel maps by gray shaded segments. A substantial shift in the lower quartile was seen for both sample network volumes dropping from 23.8 and 25.0 mmHg

under baseline conditions with normal consumption to 17.2 and 17.4 mmHg with a 40% FCD loss. The decrease in the lower quartile tissue PO_2 values for network I are clearly visualized in Figure 5 showing the fall of tissue PO_2 levels at the venous end of the network following the imposed perfusion loss and associated flow redistribution. Minimum tissue PO_2 fell to 13.5 and 15.4 mmHg compared to the baseline levels of 19.5 and 22.4 mmHg in networks I and II respectively. The lower left panel of Figure 5 shows the tissue PO_2 of network I with 40% FCD loss following RBC supply rates being restored to baseline. The lower quartile values for networks I and II with 40.0% and 42.9% FCD loss and restoration of baseline RBC supply rate increased tissue PO_2 to 22.04 and 24.5 mmHg, while minimum tissue PO_2 similarly increased to 18.9 and 22.5 mmHg respectively.

Mixed capillary outflow saturation (SvO_2) decreased progressively with increasing FCD loss. By imposing a 40.0% and 42.9% FCD loss the corresponding drop in RBC supply rate caused SvO_2 to drop to 12.9 and 13.2 percent saturated compared to 23.8 and 26.9 percent saturated under baseline conditions. With baseline RBC supply rate restored in network I at 40.0% FCD loss and network II at 42.9% FCD loss, SvO_2 increased to 23.5 and 27.4 percent saturation in the two networks respectively.

A second set of steady state oxygen transport simulations were completed, for each network volume, with the highest level of FCD loss (40.0% and 42.9%) and double the baseline oxygen consumption, at incrementally increasing RBC supply rates (Figure 4). Median tissue PO_2 levels for double oxygen consumption and baseline RBC supply rate was 11.5 mmHg for network I and 8.9 mmHg for network II. The lower quartile for baseline RBC supply rate lay at 2.7 and 3.3 mmHg with simulations resulting in minimum PO_2 of 0.0 and

0.4 mmHg for network I and II respectively. A doubling of the baseline RBC supply rate was not sufficient to restore median tissue PO₂ levels in either network with the median rising to 20.6 and 22.6, lower quartiles of 11.7 and 19.7, and minimum tissue PO₂ of 9.8 and 15.0 mmHg in networks I and II respectively. A 3.5 fold increase in RBC supply rate restored median tissue PO₂ in network II (27.9 mmHg) whereas a 4-fold increase was necessary to restore baseline levels in network I (28.0 mmHg). Increasing RBC supply rate resulted in distinctly narrowed PO₂ distributions in each network volume, particularly in the second and third quartiles.

With doubled oxygen consumption rate and baseline RBC supply rate the SvO₂ fell to 0.0 and 0.4 percent saturation in networks I and II. Doubling RBC supply rate dramatically increased SvO₂ in both networks, restoring outflow saturations to 16.9 and 23.5 percent saturation in networks I and II respectively. Higher RBC supply rates resulted in SvO₂ well above baseline levels. The 3.5 fold increase in RBC supply rate caused SvO₂ of 34.5 percent saturation in network I and 38.6 percent saturation in network II, while 4 times baseline supply rate raised outflow saturations to 44.4 and 41.3 percent saturation in the two networks respectively.

Pressures needed to create RBC flux conditions in each simulation are shown in Table 2 and 3 for networks I and II respectively. Under baseline boundary pressures increasing perfusion loss (PL) resulted in decreased RBC supply rate in each network though supply rate decreases were dependent on individual vessels that were caused to stop and therefore did not create linear decreases in supply rate with increasing FCD loss. In order to increase RBC supply rate in the high consumption (HC) simulations it was necessary to increase

boundary pressures such that the mean pressure drop increased by a factor larger than the scale of desired RBC supply rate increase.

DISCUSSION

Using real microvascular geometry, and associated experimental intravital microscopic measurements from rat extensor digitorum longus, we have shown how FCD loss with a concomitant decrease in RBC supply rate would impact tissue oxygenation. Further we have demonstrated that restoration of baseline RBC supply rate will restore original tissue PO_2 levels and that in order to compensate for increased metabolic consumption with FCD loss, it is necessary for blood flow to increase several fold. With normal consumption, the highest tested FCD loss caused up to a 27% decrease in median tissue PO_2 and a 31% decrease in minimum PO_2 . By increasing the total RBC supply rate to the simulated tissue volume under normal consumption conditions and the highest level of FCD loss, median tissue PO_2 was restored to within < 5% below baseline, and minimum tissue PO_2 was returned to within 3% of baseline in network I. In network II, restoration of baseline RBC SR caused median tissue PO_2 to be increased to within 1% of baseline. Control simulations with no stopped vessels and RBC supply rates matched to the incremental FCD loss simulations showed decrease in tissue PO_2 comparable to the corresponding level of FCD loss; this suggests that in the absence of high consumption the decreased PO_2 was primarily due to impaired supply resulting from stopped flow vessels and not from higher diffusion distance to flowing capillaries. Furthermore, the presence of anastomoses between adjacent capillaries allow for blood flow to redistribute around stopped flow vessels limiting the impact a heterogeneous stoppage of capillaries may have on oxygen delivery.

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Increased oxygen consumption in the presence of high FCD loss with baseline flow conditions caused a radical drop in tissue PO_2 creating anoxic regions in sample networks (Figure 4 & 6) and resulting in substantial hypoxia throughout the tissue volumes while reducing median tissue PO_2 to 41% (network I) and 32% (network II) of baseline levels. Under these conditions 26% of tissue elements in network I and 14% in network II had PO_2 levels below 3 mmHg. Cellular metabolism has been previously shown to decrease under hypoxic conditions in an oxygen dependent fashion when PO_2 drops below 2 mmHg [31]. However recent work by Wilson in isolated mitochondria [32] and Golub in skeletal muscle [16] found critical PO_2 (the partial pressure resulting in half maximal respiration) to be between 10 – 12 mmHg. By doubling RBC supply rate, tissue PO_2 was partially recovered in simulations for both network volumes, restoring tissue oxygen levels to 73% and 82% of baseline in networks I and II respectively and no tissue elements with PO_2 below 10 mmHg. In order to return median tissue PO_2 back to baseline under high consumption and high FCD loss conditions it was necessary to increase RBC supply rate by 4 fold in network I resulting in a tissue PO_2 99.6% of baseline while an increase of 3.5 fold was necessary in network II to restore median tissue PO_2 to 101% of baseline levels. This clearly illustrates that a substantial increase in RBC flux is necessary to compensate for the combination of observed FCD loss and the estimated high consumption present in microvascular septic injury. Control simulations that matched increases in RBC supply rates in the absence of FCD loss showed that a substantial proportion of the depressed tissue PO_2 was a result of increased diffusion distance in the ~40% FCD loss simulations indicating that perfusion loss combined with high consumption results in both a supply and diffusion limitation. The pressure increases required to raise supply rate in the high consumption simulations was 25 – 34% higher in the ~40% FCD loss cases than in controls (Table 2 and

3). This illustrates that meeting supply rate demands in the face of a large number of stopped flow vessels would require unusually high driving pressure.

In addition to tissue PO_2 we examined mixed capillary outflow saturations in each of the presented simulation cases. We found that progressive functional capillary density loss caused mixed capillary outflow saturation to decrease up to 50% compared to baseline. Decreases in mixed capillary outflow saturation can be compared with clinical findings that show a drop in mixed venous saturation in patients with severe sepsis which has also proven to be an indicator of higher patient mortality [25]. In the simulation cases where high consumption was used to simulate severe sepsis we found that increasing RBC supply rate by a factor of 2 restored median tissue PO_2 to within 18% - 27% below baseline and this was reflected in a similar recovery of SvO_2 to 14% - 29% below baseline values. However it is important to note that the blood flow increases needed to restore median tissue PO_2 to baseline levels increased SvO_2 to 40% - 87% higher than baseline levels.

Previously, Goldman et al. [12] utilized parallel capillary arrays to examine the effect of perfusion loss in sepsis and found that tissue PO_2 decreased in simulated average sepsis (tissue consumption $M_o = 3.86E-4$ mL O_2 /mL/s) to a mean value of 35.4 mmHg from 43.0 mmHg in control cases. Our findings show a PO_2 of 10.3 - 11.5 mmHg in high consumption/high perfusion loss simulations versus 28.1 - 28.2 mmHg at baseline. There are two main reasons for the differences in mean tissue PO_2 determined in this study compared to previous findings. First, Goldman et al. utilized an inlet blood saturation of 69%, which is slightly higher than our experimentally measured value of 63% for the specific networks used in our simulations. Higher inlet saturation would result in slightly elevated tissue PO_2 than what our simulations predict. Second, the relative capillary density

determined for the real networks used in the current study were 665 cap/mm² and 855 cap/mm² compared to the density chosen by Goldman et al. of 1000 cap/mm² and 1500 cap/mm². The high vascular density, blood velocities and hematocrit used to determine blood flow in the previous study result in much higher RBC supply rates than those used as the input for our simulations. Examination of our results in conjunction of Goldman's parallel arrays allows for a greater context of the delivery of oxygen *in vivo* in regions of varying vascularization and metabolic demand.

The modeling approach that we have outlined in the current work is a strong foundation for future efforts in studying the impact sepsis has on oxygen delivery as a result of microvascular injury and flow dysfunction. This approach provides a clearer picture of how perfusion loss and changes to metabolism affect tissue oxygenation, and the extent to which hypoxia may be present in tissue during a septic insult. Future improvements to the model will require precise maps of vascular geometry combined with direct measurements of individual vessels within the same geometry from septic animals *in vivo*. Previously outlined strategies for applying blood flow models to sparse hemodynamic data sets[9] combined with tracking the specific perfusion loss within a given geometry will provide a further context the current study is unable to provide. The current work represents the most realistic simulations that can be achieved without tracking loss of perfusion on a vessel-by-vessel basis, which would be the next step to fully characterize the impact of perfusion loss *in vivo*.

CONCLUSION

We have described a comprehensive and detailed model of oxygen transport that utilizes capillary networks reconstructed from intravital videos of rat skeletal muscle combined with a flow model that allows hemodynamic and oxygen saturation measurements to accurately represent conditions measured in the network *in vivo*. This model of oxygen transport simulates the blood flow dysfunction observed in skeletal muscle during experimental sepsis and reproduces varying degrees of functional capillary density loss, increases in tissue oxygen consumption, and recovery of the tissue by increasing red blood cell supply rate. Further, we have related how capillary outflow saturation changes as a result of loss of perfusion and increased consumption. Our findings provide insight into how tissue oxygenation may be affected by flow dysfunction in sepsis and that tissue hypoxemic conditions caused by a loss of vessel perfusion can be recovered by increasing total tissue perfusion to simulation volumes.

PERSPECTIVE

The functional impact of capillary perfusion loss in human disease is still difficult to characterize. Examination of this problem in models of skeletal muscle provides a meaningful analog to conditions in vital organs that is important in and of itself given the physiological importance of skeletal muscle and the relative tissue mass that is potentially affected. While predictions for mixed capillary outflow saturation do not translate directly to central venous measurements made in patients, it does provide a tangible reference point that puts into context how changing vascular density, PO_2 , and consumption in small vascular volumes might impact global measurements. While this work does elucidate how

loss of perfused capillaries would impact tissue PO₂, the next step is to repeat our approach using data collected directly from septic animals, tracking the specific vessel loss, and directly calculating changes in consumption within observed discrete volumes. Collection of such a detailed data set, particularly from diseased animals, will remain a challenge. This will be a critical step in order to move this work forward and serves as strong motivation to carry out 3D reconstructions and modeling directly from an animal model of sepsis.

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Table 1: Summary of network geometry parameters.

Statistical summary of network geometry data.

	Tissue Volume Dimensions ($\mu\text{m} \times \mu\text{m} \times \mu\text{m}$)	Number of Vessel Segments	Mean Capillary Segment Length (μm)	Capillary Length Density ($\text{mm} \times \text{mm}^3$)
Network I	84 x 168 x 342	25	107.4 \pm 82.7	554.2
Network II	70 x 157 x 268	20	100.9 \pm 74.7	684.8

Table 2. Variables for Network II simulations outlining the four different levels of perfusion loss; high consumption simulations are indicated with associated increasing levels of RBC supply rate ranging from baseline RBC supply rate to 4X times baseline RBC flux.

Network I Simulations	% FCD Loss	Consumption - M_0 (mL O ₂ /mL tissue/s)	RBC Supply Rate (mL RBC/mL tissue/s)	Mean Pressure Drop (mmHg)
Baseline	0	1.5E-4	8.25E-04	3.00
6.7% FCD Loss	6.7	1.5E-4	7.67E-04	3.00
Control, SR Matched to 6.7% FCD Loss	0	1.5E-4	7.67E-04	2.80
20% FCD Loss	20	1.5E-4	7.43E-04	3.00
Control, SR Matched to 20% FCD Loss	0	1.5E-4	7.43E-04	2.71
26.7 FCD Loss	26.7	1.5E-4	6.83E-04	3.00
Control, SR Matched to 26.7% FCD Loss	0	1.5E-4	6.83E-04	2.50
40% FCD Loss	40	1.5E-4	6.07E-04	3.00
Control, SR Matched to 40% FCD Loss	0	1.5E-4	6.07E-04	2.23
40% FCD Loss, Baseline SR	40	1.5E-4	8.25E-04	4.03
40% FCD Loss, High Consumption, Baseline SR	40	3.0E-4	8.25E-04	4.03
Control, 0% FCD Loss, High Consumption, Baseline SR	0	3.0E-4	8.25E-04	3.00
40% FCD Loss, High Consumption, 2X SR	40	3.0E-4	1.65E-03	7.93
Control, 0% FCD Loss, High Consumption, 2X SR	0	3.0E-4	1.65E-03	5.92
40% FCD Loss, High Consumption, 3.5X SR	40	3.0E-4	2.89E-03	13.74
Control, 0% FCD Loss, High Consumption, 3.5 SR	0	3.0E-4	2.89E-03	10.29
40% FCD Loss, High Consumption, 4X	40	3.0E-4	3.30E-03	15.69

SR				
Control, 0% FCD Loss, High Consumption, 4X SR	0	3.0E-4	3.30E-03	11.75

Table 3. Variables for Network II simulations outlining the four different levels of perfusion loss; high consumption simulations are indicated with associated increasing levels of RBC supply rate ranging from baseline RBC supply rate to 4X times baseline RBC flux.

Network II Simulations	% FCD Loss	Consumption - M_0 (mL O ₂ /mL tissue/s)	RBC Supply Rate (mL RBC/mL tissue/s)	Mean Pressure Drop (mmHg)
Baseline	0	1.5E-4	6.84E-04	1.98
7.1% FCD Loss	7.1	1.5E-4	6.62E-04	1.98
Control, SR Matched to 7.1% FCD Loss	0	1.5E-4	6.62E-04	1.92
21.4% FCD Loss	21.4	1.5E-4	6.12E-04	1.98
Control, SR Matched to 21.4% FCD Loss	0	1.5E-4	6.12E-04	1.78
28.6 FCD Loss	28.6	1.5E-4	5.91E-04	1.98
Control, SR Matched to 28.6% FCD Loss	0	1.5E-4	5.91E-04	1.72
42.9% FCD Loss	42.9	1.5E-4	4.37E-04	1.98
Control, SR Matched to 42.9% FCD Loss	0	1.5E-4	4.37E-04	1.30
42.9% FCD Loss, Baseline SR	42.9	1.5E-4	6.84E-04	2.96
42.9% FCD Loss, High Consumption, Baseline SR	42.9	3.0E-4	6.84E-04	2.96
Control, 0% FCD Loss, High Consumption, Baseline SR	0	3.0E-4	6.84E-04	1.98

42.9% FCD Loss, High Consumption, 2X SR	42.9	3.0E-4	1.37E-03	5.70
Control, 0% FCD Loss, High Consumption, 2X SR	0	3.0E-4	1.37E-03	3.88
42.9% FCD Loss, High Consumption, 3.5X SR	42.9	3.0E-4	2.39E-03	8.41
Control, 0% FCD Loss, High Consumption, 3.5 SR	0	3.0E-4	2.39E-03	6.72
42.9% FCD Loss, High Consumption, 4X SR	42.9	3.0E-4	2.74E-03	11.13
Control, 0% FCD Loss, High Consumption, 4X SR	0	3.0E-4	2.74E-03	7.67

Table 4. List of constants and values used in oxygen transport simulations.

Constant	Value
α	3.89 E-5 ml O ₂ ml ⁻¹ mmHg ⁻¹
α_b	2.82 E-5 ml O ₂ ml ⁻¹ mmHg ⁻¹
C	0.52 ml O ₂ ml ⁻¹
D	2.41 E-5 cm ² s ⁻¹
P_{cr}	0.5 mmHg
C_{Mb}	1.02 E-2 ml O ₂ ml ⁻¹
D_{Mb}	3 E-7 cm ² s ⁻¹
P_{50}	37 mmHg
$P_{50,Mb}$	5.3 mmHg

















