Development of a species-specific model of cerebral hemodynamics

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In this paper, a mathematical model for the description of cerebral hemodynamics is developed. This model is able to simulate the regulation mechanisms working on the small cerebral arteries and arterioles, and thus to adapt dynamically the blood flow in brain. Special interest is laid on the release of catecholamines and their effect on heart frequency, cardiac output and blood pressure. Therefore, this model is able to describe situations of severe head injuries in a very realistic way.

Keywords: Cerebral hemodynamics; Regulation mechanisms; Catecholamines; Cardiac output

1. Introduction

Cerebral perfusion is a critical parameter in many clinical situations, e.g. cerebral infarction or head injury. Many of these conditions have been investigated extensively in experimental and clinical studies with respect to a wide variety of clinical and physiological parameters. In the past decades, the key mechanisms involved in the regulation of cerebral perfusion have been identified. However, it is a well-known problem of classical reductionist-experimental approaches that different physiological parameters cannot be evaluated simultaneously. Consequently, there is very little knowledge concerning the in-situ consequences of interactions between different physiological regulatory systems. The therapeutic interventions based on single parameter mechanisms, e.g. hyperventilation to reduce arterial $pCO_2$, have not proven successful as “single-agent” therapy to improve the outcome in head injured patients. Physiologic data from patients in a clinical context are difficult to obtain as these patients are often in a life threatening condition. Given the high degree of susceptibility of traumatic brain injuries to any external stimuli extensive interventions for data acquisition are ethically questionable as these interventions might affect the outcome. Data collection in healthy control groups is even more problematic with respect to possible complications due to the highly invasive nature of measurement technology. Additionally, the varying extend of traumatic brain injury, comorbidities or additional injuries result in very heterogeneous patient populations, i.e. the data pool for modelling will be the sum total of very different clinical entities that might display differing behaviour. Mathematical approaches do not suffer from these intrinsic problems. In the past different mathematical models of the cerebral circulation or distinct parts of the regulatory systems have been developed. However, the two major drawbacks of these models are a lack of anatomical coherence and the missing species specificity. For these reasons it has been difficult to validate these models experimentally. Consequently, the ultimate aim of all modelling approaches in clinical medicine, i.e. to arrive at a level of understanding of physiological processes sufficiently profound to derive rules to influence the in-vivo system systematically (i.e. therapeutically), has not yet been reached. The model presented in this paper is, therefore, strictly species specific as almost exclusively data from experiments with Sprague-Dawley rats have been used for parameter estimation. To facilitate the experimental validation, as well as model based experimental interventions the structure of the model is kept very close to the anatomical structure in-vivo.

The basic idea of this model is the treatment of blood flow through extra- and intracranial vessels as a hydraulic circuit. This is a standard way to describe blood flow dynamics as can be found in references [1–4]. The advantage of this approach is the portability of the fundamental laws of electric circuits to hydraulic circuits, like Ohm’s and Kirchhoff’s law.
2. Qualitative model description

The model is qualitatively presented, with attention focused on its new aspects.

2.1 Extracranial arterial pathways

The work of Ursino et al. [2] is used as the basic model for the new investigations. There only the cerebral hemodynamics are considered and the blood pressure $P_a$ is chosen as a constant input parameter for the cerebral blood circulation. In this work, the extracranial arterial pathways are also modelled and thus the arterial blood pressure is no constant parameter but depends on time $t$, cardiac output $Q$ and thus on cardiac parameters.

The segment of the extracranial arteries from the left heart to the large cerebral arteries is, like the other segments of the model, described by the hydraulic resistance $R_a$ and the hydraulic compliance $C_a$. The amount of blood ejected from the heart into the aorta in a certain time is modelled by a function $Q$, which will be described later. Because of the theory of hydraulic circuits, blood volume changes $dV/dt$ in the extracranial and intracranial arteries and veins are given by the difference of blood flow into and out of these vessels. The compliances $C$ are synonymous with the storage capacities of the arteries and veins. Therefore, the blood volume changes $dV/dt$ are given by the difference of the blood flow into the aorta ($Q$) minus the systemic blood flow out of the aorta through all organs of the body ($(P_a - P_{cv})/R_s$):

$$\frac{dV_a}{dt} = C_a \frac{dP_a}{dt} = Q - \frac{P_a - P_{cv}}{R_s} \quad (1)$$

where $Q$ is cardiac output, $P_a$ arterial, $P_{cv}$ central venous pressure and $R_s$ systemic resistance. The blood flow through the vessels is calculated by Ohm’s law. Since $P_{cv} \ll P_a$ changes in blood pressure $dP_a/dt$ are approximately given by

$$\frac{dP_a}{dt} = \frac{1}{C_a} \left( Q - \frac{P_a}{R_a} \right). \quad (2)$$

The fraction of cardiac output $Q$ which goes into head is then given by $(P_a - P_{ls})/R_{ls}$ and the value of the compliance $C_a$ is corresponding to [5] $C_a = 0.0042$ ml/mm Hg.

2.2 Cardiac output

The model function for cardiac output $Q$, developed by Stevens et al. [6], is used to get a pulsatile blood flow throughout the circulatory system. The cardiac output $Q$ is modelled by defining an interior function which oscillates with the frequency of the heart pulse and an envelope function for these interior oscillations. The product of these two functions is then normalized and the parameters are calibrated with physical rat data. The provisional flow function $Q_3(t,n,\Phi)$ is then given by

$$Q_3(t,n,\Phi) = Q_1(t,n) \cdot Q_2(t,\Phi), \quad (3)$$
where the envelope function is defined by

$$Q_1(t, n) = \sin^n(\omega t) \text{ with } n \text{ odd}$$

(4)

and the interior function has the form

$$Q_2(t, \Phi) = \cos(\omega t - \Phi)$$

(5)

with $\omega$ one half of the basic frequency of the heart pulse and $\Phi$ a suitable phase angle. Generally $\Phi$ lies in the range $0 < \Phi \leq \pi/2$. If $\Phi = 0$, cardiac outflow will equal backflow. There is nearly zero backflow if $\Phi > \pi/6$. For $n = 13$ and $\Phi = 0$, these two functions are shown in figure 2.

The $Q$ must be normalized and calibrated to produce a good model function for cardiac output. The set of calibration parameters for this model function includes the stroke volume $\nu$, the heart rate $b$, and the phase angle $\Phi$. To fit the experimental data given by reference [5], the parameters are chosen as follows:

the heart rate $b = 378/60$ beats per second, the mean value for cardiac output $Q = \nu b = 70/60 \text{ ml s}^{-1}$ thus the stroke volume $\nu = Q/b = 0.1852 \text{ ml per second}$ the phase angle $\Phi = \pi/10$.

Once appropriate values of these calibration parameters are chosen, it is possible to determine the period $p = 1/b$ of the cycle and the frequency $\omega = \pi/p$.

By normalizing the model function $Q_3$ so that the total outflow over one period equals $\nu$ one gets

$$Q(t, n, \Phi) = \frac{\nu}{\tilde{A}(n, \Phi)} Q_3(t, n, \Phi)$$

$$= \frac{\nu}{\tilde{A}(n, \Phi)} \sin^n(\omega t) \cos(\omega t - \Phi)$$

(6)

where

$$\tilde{A}(n, \Phi) = \int_0^p Q_3(t, n, \Phi) dt.$$  

(7)

With the relation $\omega = \pi/p$ and noting that

$$\tilde{A}(n, \Phi) = \int_0^\pi \sin^n(\omega t) \cos(t - \Phi) dt$$

$$= \frac{\sqrt{\pi} \Gamma\left(1 + \frac{n}{2}\right) \sin(\Phi)}{\Gamma\left(\frac{1+n}{2}\right)}$$

(8)

with $\Gamma$ the Euler gamma function, one gets

$$A(n, \Phi) = \int_0^p Q_3(t, n, \Phi) dt$$

$$= \int_0^\frac{\pi}{\omega} \sin^n(\omega t) \cos(\omega t - \Phi) dt = \frac{\tilde{A}(n, \Phi)}{\omega}.$$  

(9)

An example for $Q(t, n, \Phi)$ is given in figure 3. The narrowness of the output function $Q$ is determined by the choice of $n$. A large $n$ represents a small systole period. Using a value of $n = 13$ results in a systole period approximately $1/3$ of the cardiac cycle, consistent with values given by many of the standard texts in physiology.
2.3 Intracranial hemodynamics

The Monro–Kellie doctrine implies that any volume variation in an intracranial compartment causes a compression or dislocation of the other volumes. These changes in the compartments are accompanied by an alteration in intracranial pressure $P_{ic}$. The intracranial compliance $C_{ic}$, which represents the capacity of the craniospinal system to store a volume load, is according to Ursino et al. [2] assumed to be inversely proportional to intracranial pressure through a constant parameter

$$C_{ic} = \frac{1}{k_E P_{ic}}. \quad (10)$$

In this model, volume changes in the craniospinal space are ascribed to four compartments: large and middle cerebral arteries $dV_{la}/dt$, pial arteries and arterioles $dV_{pa}/dt$, cerebral veins $dV_{v}/dt$, and the H$_2$O compartment $dV_{H_2O}/dt$. According to the Monro–Kellie doctrine the following conservation equation holds

$$C_{ic} \frac{dP_{ic}}{dt} = \frac{dV_{la}}{dt} + \frac{dV_{pa}}{dt} + \frac{dV_{v}}{dt} + \frac{dV_{H_2O}}{dt} \quad (11)$$

with time $t$. The biomechanical analog in figure 1 represents the four intracranial compartments considered in the model, together with the extracranial arterial and venous pathways.

2.4 Large and middle cerebral arteries

The first intracranial segment of the model represents the circulation of blood in the large and middle cerebral arteries. The hemodynamic is described by a hydraulic resistance $R_{la}$ and a hydraulic compliance $C_{la}$. In contrast to the model of Ursino et al. [2] changes in the storage capacity $C_{la}$ and thus on the blood volume $V_{la}$ and the pressure $P_{la}$ are modelled. The changes in volume in this segment $dV_{la}/dt$ are given by

$$\frac{dV_{la}}{dt} = \frac{P_{la} - P_{pa} - P_{pa} - P_{pa}}{R_{la} - R_{pa}/2} \quad (12)$$

where $P_{pa}$ and $R_{pa}$ are pressure and resistance of the pial arteries and arterioles, respectively.

Because the impact of the cerebrovascular regulation mechanisms on these intracranial arteries is very small, the resistance $R_{la}$ is assumed to be constant. Further on, volume changes in this compartment depend only on changes in transmural pressure $(P_{la} - P_{pa})$ and not on changes in compliance $C_{la}$, since these vessels behave passively. Thus the following equation holds

$$\frac{dV_{la}}{dt} = C_{la} \left( \frac{dP_{la}}{dt} - \frac{dP_{ic}}{dt} \right) \quad (13)$$

With these two equations in mind one gets a differential equation which describes pressure changes $dP_{la}/dt$ in the large and middle cerebral arteries:

$$\frac{dP_{la}}{dt} = \frac{1}{C_{la}} \left( \frac{P_a - P_{la}}{R_{la}} - \frac{P_{la} - P_{pa}}{R_{pa}/2} \right) + \frac{dP_{ic}}{dt}. \quad (14)$$

The compliance of these vessels is assumed to be inversely proportional to the transmural pressure

$$C_{la} = \frac{k_{C_{la}}}{P_{la} - P_{ic}} \quad (15)$$

with $k_{C_{la}}$ the proportionality constant.

2.5 Pial arteries and arterioles

In this compartment of the model all sections of the cerebrovascular bed directly under the control of the regulatory mechanisms are comprised. This pial arterial segment is described by a hydraulic resistance $R_{pa}$ and a hydraulic compliance $C_{pa}$. Both of these parameters are regulated by cerebrovascular control mechanisms. The two equations which describe the changes in volume $dV_{pa}/dt$ in this segment and the calculation of the pressure at the cerebral capillaries $P_{c}$ (applying Kirchhoff’s law) are given in reference [2].

With these three equations the pressure change $dP_{pa}/dt$ in the pial arterial compartment is described by

$$\frac{dP_{pa}}{dt} = \frac{1}{C_{pa}} \left( \frac{P_{la} - P_{pa}}{R_{pa}/2} - \frac{P_{pa} - P_{c}}{R_{pa}/2} - \frac{dC_{pa}}{dt} (P_{pa} - P_{ic}) \right) + \frac{dP_{ic}}{dt}. \quad (16)$$

2.6 Intracranial and extracranial venous circulation

The intracranial vascular bed of the veins is described by a series arrangement of two segments. The first, from the small postcapillary venules to the large cerebral veins, contains the resistance $R_{pv}$ and the venous compliance $C_{vi}$. Corresponding to reference [2], the compliance is calculated by

$$C_{vi} = \frac{k_{ven}}{P_{v} - P_{ic} - P_{vi}}, \quad (17)$$

where $k_{ven}$ is a constant parameter and $P_{vl}$ represents the transmural pressure value at which cerebral veins collapse.

Using the equations defined in reference [2], which describe the volume changes $dV_{v}/dt$ of this venous compartment, the pressure changes $dP_{v}/dt$ are given by

$$\frac{dP_{v}}{dt} = \frac{1}{C_{vi}} \left( \frac{P_{c} - P_{v}}{R_{pv}} - \frac{P_{v} - P_{v}}{R_{ven}} \right) + \frac{dP_{ic}}{dt}. \quad (18)$$

where $R_{ven}$ is the resistance of the terminal intracranial veins and $P_{vi}$ the pressure at the dural sinuses.

The second segment represents the terminal intracranial veins (e.g. lateral lakes). During intracranial hypertension
these vessels collide or narrow at their entrance into the dural sinuses, with a mechanism similar to that of a starling resistor (cf. [2]). Because of this phenomenon the resistance $R_{vs}$ depends on the pressures of the system in the following way:

$$R_{vs} = \frac{P_v - P_{vs}}{P_v - P_{ic}} R_{vs1},$$

where $R_{vs1}$ represents the terminal vein resistance when $P_{ic} = P_{vs}$.

In contrast to the model of Ursino et al. [2] the sinus venous pressure $P_{vs}$ is not assumed to be constant, but depends on time and the other pressures of the system and is calculated by Kirchhoff’s law

$$\frac{P_v - P_{vs}}{R_{vs}} + q_o = \frac{P_{vs} - P_{cv}}{R_{ve}}.$$  \hspace{1cm} (20)

Since the water backflow at the dural sinuses $q_o$ is negligible in comparison to the blood flows, it is assumed to be zero.

The extracranial venous circulation from the dural sinuses through the vena cava back to the heart is described by the hydraulic resistance $R_{ve}$ and the hydraulic compliance $C_{ve}$. Because no mechanisms acting on these blood vessels are taken into account, these parameters are assumed to be constant.

### 2.7 H₂O compartment

Under clinical aspects the formation of cerebral edema after head injury has to be described by the model. This mechanism is reproduced by water outflow at the capillaries into the craniospinal space and water backflow at the dural sinuses. It is assumed that the two processes are passive and unidirectional, thus the following equations hold:

$$q_f = \begin{cases} \frac{P_v - P_{ic}}{R_f} & \text{if } P_{ic} > P_{ic} \\ 0 & \text{else} \end{cases}$$ \hspace{1cm} (21)

$$q_o = \begin{cases} \frac{P_v - P_{vs}}{R_o} & \text{if } P_{ic} > P_{vs} \\ 0 & \text{else.} \end{cases}$$ \hspace{1cm} (22)

The case of a severe cerebral edema is simulated by decreasing the outflow resistance $R_f$, thus increasing the outflow $q_f$, whereas the backflow $q_o$ is assumed to be constant and small all the time. Under physiological conditions $q_f$ and $q_o$ are approximately zero. Changes in volume in this H₂O compartment are given by $dV_{H₂O}/dt = q_f - q_o$.

### 2.8 Cerebrovascular regulation mechanisms

Cerebrovascular regulation mechanisms work by modifying the resistance $R_{pa}$ and the compliance $C_{pa}$ (and hence the blood volume) in the pial arterial–arteriolar vasculature.

In this section, three mechanisms are considered which regulate cerebral blood flow. The effects of two of them, like autoregulation and CO₂ reactivity, are described in [2]. One new cerebrovascular regulation mechanism, the NO reactivity, is inserted into the model and its effect on the pial arterial compliance is modelled by using the given idea of a sigmoidal relationship of the whole regulation process.

#### 2.8.1 Autoregulation

The cerebral autoregulation describes the ability of certain vessels to keep the cerebral blood flow (CBF) relatively constant despite changes in perfusion pressure.

As you can see in the upper branch of figure 4, it is assumed that autoregulation is activated by changes in CBF. The impact of this mechanism on the pial arterial vessels is described by a first-order low-pass filter dynamic with time constant $\tau_{aut}$ and gain $G_{aut}$ (cf. [2])

$$\tau_{aut} \frac{dx_{aut}}{dt} = -x_{aut} + G_{aut} \left( q - q_n \right),$$ \hspace{1cm} (23)

where $q$ is the measured CBF and $q_n$ the cerebral blood flow under basal conditions.

The cerebral blood flow $q$ can be calculated by Ohm’s law

$$q = \frac{P_{pa} - P_{a}}{R_{pa}/2}.$$ \hspace{1cm} (24)

With this relation, we get a basal value for blood flow through the pial arteries of $q_n = 0.1696 \text{ ml s}^{-1}$. The value of the gain $G_{aut}$ is given by fitting the autoregulation curve of [7].

#### 2.8.2 CO₂ reactivity

The CO₂ reactivity describes the dependence of cerebral blood flow on arterial CO₂ pressure $P_{aCO₂}$.

The branch in the middle of figure 4 represents the CO₂ reactivity, which is activated by changes in $P_{aCO₂}$ and described by a first-order low-pass filter dynamic with time constant $\tau_{CO₂}$ and gain $G_{CO₂}$ (cf. [2])

$$\tau_{CO₂} \frac{dx_{CO₂}}{dt} = -x_{CO₂} + G_{CO₂} A_{CO₂} \log_10 \left( \frac{P_{aCO₂}}{P_{aCO₂}} \right),$$ \hspace{1cm} (25)

where $P_{aCO₂}$ is the CO₂ pressure under basal conditions, corresponding to [8] it is $P_{aCO₂} = 33 \text{ mm Hg}$. $A_{CO₂}$ is a corrective factor, which will be described later. The value of the gain $G_{CO₂}$ is obtained by fitting the data of Iadecola et al. [9].

#### 2.8.3 NO reactivity

The NO reactivity describes the dependence of cerebral blood flow on the production rate of NO at the endothelial cells of the pial vessels $q_{NO}$.
For this regulation mechanism the following assumptions are made: first, only the impact of nitric oxide (NO) on the smooth muscle cells of pial vessels is considered, whereas the response of large arteries and veins on nitric oxide is neglected. Second, although there are distinct sources of NO in brain, e.g. neuronal or endothelial NO, the model does not differentiate the different sources of NO. Furthermore, no interactions of NO with other substances are considered.

In the case of a head injury production of nitric oxide occurs at the endothelial cells of the pial arteries and arterioles. These NO molecules migrate through the vessel wall to the smooth muscle cells and activate a substance called guanylyl cyclase there, which causes a higher production of guanosine 3',5'-cyclic monophosphate (cGMP) with subsequent relaxation. In contrast any decrease in NO production causes constriction of the pial vessels.

The lower branch of figure 4 represents the NO reactivity, which is activated by changes in the NO production rate \( q_{NO} \) and described by a first-order low-pass filter dynamic with time constant \( \tau_{NO} \) and gain \( G_{NO} \)

\[
\tau_{NO} \frac{dx_{NO}}{dt} = -x_{NO} + G_{NO} \log_{10} \left( \frac{q_{NO}}{q_{NON}} \right),
\]

where \( q_{NON} \) defines the NO production rate under basal conditions, corresponding to [10] it is \( q_{NON} = 54.1 \) ng/g tissue.

It is assumed that the production rate \( q_{NO} \) is linearly correlated with the concentration of nitric oxide \( C_{NO} \) in the vessel wall and that the vessel radius dependence on \( \log_{10} \) of NO concentration is almost linear in the range of physiological conditions (cf. [11,12]). These are the reasons why the logarithm of \( q_{NO} \) is chosen as input to the controller. The regulation gain \( G_{NO} \) is then given by fitting the data of Wang et al. [13].

The time constants of these three regulation mechanisms \( \tau_{aut} \), \( \tau_{CO2} \) and \( \tau_{NO} \) are approximated by using data of references [9,13,14], the values are 20, 50 and 40 s, respectively.

The minus sign at the upper branch of figure 4 means that an increase in cerebral blood flow causes vasodilation, with a decrease in pial compliance and an increase in pial resistance as consequence. Whereas, the plus signs at the middle and lower branches of figure 4, mean that an increase in arterial CO2 pressure or in endothelial NO production causes vasodilation, with an increase in pial compliance and a decrease in pial resistance as consequence.

2.9 Impact of the regulation mechanisms on \( C_{pa} \) and \( R_{pa} \) and their interactions

According to [15], the impact of NO production at the endothelial cells on arterial CO2 pressure is very small. Therefore, only the effect of \( P_{aCO2} \) on the production rate \( q_{NO} \) is considered in this model. According to [16], an 70% increase in arterial CO2 pressure yields an 20% increase in the NO production rate. It is assumed that these two mechanisms are connected linearly by the following equation

\[
q_{NO} = 0.4332 P_{aCO2} + 39.8048.
\]
of CO₂ changes on perivascular pH. This phenomenon is modelled by the corrective factor $A_{CO₂}$.

$$A_{CO₂} = \frac{1}{1 + \exp \left\{ - k_{CO₂} (q - q_a)/q_a \right\} - b_{CO₂} }$$

(28)

with constant parameters $k_{CO₂}$ and $b_{CO₂}$ (cf. [2]).

Another nonlinearity is given by the fact that the whole regulation process is not just the sum of these three mechanisms but is described by a sigmoidal relationship with upper and lower saturation levels. Adapting the situation of Ursino et al. [2] to three regulation mechanisms one gets

$$C_{pa} = \frac{(C_{pan} - \Delta C_{pa}/2) + (C_{pan} + \Delta C_{pa}/2) \cdot \exp \left\{ (x_{CO₂} + x_{NO} - x_{aut})/k_{Cp} \right\} }{1 + \exp \left\{ (x_{CO₂} + x_{NO} - x_{aut})/k_{Cp} \right\} }$$

(29)

where $C_{pan}$ is the pial arterial compliance under basal conditions, $\Delta C_{pa}$ the change in compliance and $k_{Cp}$ a constant parameter.

This equation shows that any decrease in CBF, any decrease in arterial CO₂ pressure and any decrease in NO production rate causes vasodilation with an increase in pial arterial compliance $C_{pa}$. On the other hand, any increase in CBF, any increase in arterial CO₂ pressure and any increase in NO production rate causes vasoconstriction with a reduction in compliance $C_{pa}$.

A value for the constant parameter $k_{Cp}$ was given to set the central slope of the sigmoidal curve to +1. This condition is obtained by assuming $k_{Cp} = \Delta C_{pa}/4$.

An important point is that this sigmoidal curve is not symmetrical: the increase in blood volume caused by vasodilation is greater than the decrease of blood volume caused by vasoconstriction. That is the reason why two different values of the parameter $\Delta C_{pa}$ have to be chosen depending on whether vasodilation or vasoconstriction is considered. It is

$$\begin{align*}
    \{ x_{CO₂} + x_{NO} - x_{aut} \geq 0 : \ & \Delta C_{pa} = \Delta C_{pa1}; k_{Cp} = \Delta C_{pa1}/4, \\
    \{ x_{CO₂} + x_{NO} - x_{aut} < 0 : \ & \Delta C_{pa} = \Delta C_{pa2}; k_{Cp} = \Delta C_{pa2}/4. 
\end{align*}$$

(30)

Consider equation (29): For $(x_{CO₂} + x_{NO} - x_{aut}) \to \infty$ one gets $C_{pa} \to (C_{pan} + \Delta C_{pa}/2)$ and $(x_{CO₂} + x_{NO} - x_{aut}) \to -\infty$ yields $C_{pa} \to (C_{pan} - \Delta C_{pa}/2)$. That means $(C_{pan} + \Delta C_{pa}/2)$ and $(C_{pan} - \Delta C_{pa}/2)$ are the upper and lower saturation levels of the sigmoidal curve, respectively.

An expression for $dC_{pa}/dr$ is obtained by differentiating equation (29) to

$$dC_{pa}/dr = \frac{\Delta C_{pa}}{k_{Cp}} \cdot \frac{\exp \left\{ (x_{CO₂} + x_{NO} - x_{aut})/k_{Cp} \right\} }{\left[ 1 + \exp \left\{ (x_{CO₂} + x_{NO} - x_{aut})/k_{Cp} \right\} \right]^2} \times \frac{d(x_{CO₂} + x_{NO} - x_{aut})}{dr}.$$ 

(31)

The cerebrovascular control mechanisms act also on the hydraulic pial arterial resistance $R_{pa}$. Because the blood volume is directly proportional to the inner radius second power, while the resistance is inversely proportional to inner radius fourth power, the following relationship holds between the pial arterial volume and resistance (cf. [2])

$$R_{pa} = \frac{k_R C_{pan}^2}{V_{pa}^2}$$

(32)

where $k_R$ is a constant parameter.

### 2.10 Norepinephrine and its impact on heart rate

Norepinephrine is the principal mediator of the sympathetic nervous system. Cardiac function is modulated in many aspects by norepinephrine. The primary effect of this substance is an increase in heart rate and thus an increase in cardiac output $Q$ (figure 5).

The changes of the norepinephrine concentration in blood $d[NE]/dr$ are described by the equation

$$\frac{d[NE]}{dr} = r - \alpha_{NE}[NE],$$

(33)

where $r$ is the constant NE release during sympathetic nerve stimulation and $\alpha_{NE}$ is the elimination rate.

Because absolute values of $[NE]$ are unknown $r$ can be fixed to one in the case of sympathetic nerve activation without loss of generality, otherwise $r$ is chosen as zero. That means this mechanism of NE release is switched on and off by the parameter $r$ and the parameter $\alpha_{NE}$ specifies the strength of sympathetic nerve stimulation and can be varied throughout the simulations.

![Figure 5. Dependence of heart rate variation hr on the amount of norepinephrine in blood. Model results (solid curve) and measured data of Muchitsch et al. ([17]).](image-url)
The heart rate response to a steplike increase of the norepinephrine concentration [NE] is described by a first-order low-pass filter dynamic with time constant \( \tau_h \)

\[
\frac{dhr}{dt} = -hr + G([NE]),
\]

where \( hr \) is the heart rate variation. The steady-state heart rate response \( G([NE]) \) is, according to [18], defined by

\[
\Delta HR = G([NE]) = \frac{\Delta HR_{max}[NE]^2}{K_{NE} + [NE]^2}, \tag{35}
\]

where \( K_{NE} \) is the NE concentration producing a half maximum response and \( \Delta HR_{max} \) is the maximum value of \( \Delta HR \). Corresponding to Muchitsch et al. [17], the values of these parameters are 175 bpm and 100 \( \mu \)g kg\(^{-1}\) min\(^{-1}\), respectively and the time constant \( \tau_h \) is 5 s.

The new heart rate \( \dot{b} \) is then given by adding the heart rate variation \( hr \) to \( b \)

\[
\dot{b} = b + hr.
\]

3. Parameter estimation

The estimation of systemic parameters under basal conditions is described now. The values of the compliances in the craniospinal space, like \( C_{in}, C_{pa}, \) and \( C_{vi} \), and the intracranial compliance \( C_{ic} \) were fitted by using pressure curves of these compartments. The values were fitted in the way that the model amplitudes of the pressures in each compartment are equal to the given physiological amplitudes of pressures (cf. [19–22]).

The basal value of the resistance of the large intracranial arteries is calculated by using the Hagen–Poiseuille law

\[
R = \frac{8\eta l}{\pi r^4},
\]

All other model resistances, \( R_a, R_{pa}, R_{pv}, R_{vs} \) and \( R_{ve} \) are calculated by using the mean pressure values in each compartment (see [19–21]) and solving the differential equations defined above. All model parameters under basal conditions are given in table 1.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Value</th>
</tr>
</thead>
<tbody>
<tr>
<td>( C_a )</td>
<td>0.0042 ml/mm Hg</td>
</tr>
<tr>
<td>( n )</td>
<td>13</td>
</tr>
<tr>
<td>( v )</td>
<td>0.1852 ml per beat</td>
</tr>
<tr>
<td>( R_{in} )</td>
<td>47.1609 mm Hg \cdot s \cdot ml(^{-1})</td>
</tr>
<tr>
<td>( \Delta C_{pa} )</td>
<td>4.7277e - 07 ml/mm Hg</td>
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<tr>
<td>( R_{pa} )</td>
<td>2830 mm Hg \cdot s \cdot ml(^{-1})</td>
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<tr>
<td>( b_{CO2} )</td>
<td>54.1 ng/g tissue</td>
</tr>
<tr>
<td>( \beta_{CO2} )</td>
<td>19</td>
</tr>
<tr>
<td>( P_{CO2} )</td>
<td>1.7 mm Hg</td>
</tr>
<tr>
<td>( P_{O2} )</td>
<td>-2.5 mm Hg</td>
</tr>
<tr>
<td>( k_E )</td>
<td>41 ml \cdot ml(^{-1})</td>
</tr>
<tr>
<td>( \tau_{hr} )</td>
<td>5 s</td>
</tr>
<tr>
<td>( R_{ic} )</td>
<td>9.4935e - 08 ml</td>
</tr>
<tr>
<td>( R_{ao} )</td>
<td>5.566 mm Hg \cdot s \cdot ml(^{-1})</td>
</tr>
<tr>
<td>( D_{max} )</td>
<td>175/60 beats per second</td>
</tr>
<tr>
<td>( k_{ne} )</td>
<td>100 ( \mu )g \cdot kg(^{-1}) \cdot min(^{-1})</td>
</tr>
</tbody>
</table>

4. Numerical simulations

With the parameters given in table 1 numerical simulations were performed to show that the model gives a reasonable and realistic description of the physiologic system.

Figure 6 shows the simulated pressure in the aorta \( P_a \) and the intracranial pressure \( P_{ic} \). These pressures agree with experimental data of Baumbach [20], who measured a systolic blood pressure of 134 ± 7 mm Hg and a diastolic pressure of 98 ± 6 mm Hg and with data of Holtzer et al. [22] who measured a mean intracranial pressure of 6 ± 3 mm Hg and an amplitude of the pressure curve of approximately 1.4 mm Hg.

The simulated pressure of the small arteries and arterioles in the left of figure 7 agrees with the measured data of 67 ± 4 mm Hg for the pial systolic pressure and 53 ± 3 mm Hg for the pial diastolic pressure by Baumbach [20]. Further on, the works of Gotoh et al. [19] and Sugiyama et al. [21] suggest a mean large cerebral arteries pressure of 108 mm Hg and an amplitude of 23 mm Hg, which agrees also with the simulated pressure curve in the right hand side part of the figure.
Figure 8 shows the simulated pressure of the cerebral veins $P_v$. A mean cerebral venous pressure of $7 \pm 1$ mm Hg is given by Mayhan and Heistad [23].

The simulated dependence of cerebral blood flow on arterial CO2 pressure is shown in the left hand side part of the figure 9. One observes a 230% increase in cerebral blood flow, if $P_{aCO_2}$ is changed from 34.3 to 49.2 mm Hg. This result corresponds to the measured data of Iadecola et al. [9]. Further on, the impact of NO on CBF is simulated and the result is given in the right of figure 9.

To show that the interaction between CO2 and NO reactivity is modelled realistically, the dependence of $P_{aCO_2}$ on NO is simulated by calculating the CO2 reactivity with the basal value of $q_{NO}$ and with inhibition of NO production. The results are shown in figure 10: the solid curve is equal to the left curve in figure 9, since $q_{NO} = q_{NO_B}$ and the dashed curve is the simulation result with inhibited NO production, $q_{NO} = 0.1 \cdot q_{NO_B}$. The resulting decrease in CBF corresponds to the data of Wang et al. [13].

In addition to the numerical simulations described above, the impact of the sympathetic system, i.e. norepinephrine on heart rate and cardiac output, is simulated. The strength of sympathetic nerve stimulation is regulated by the parameter $\alpha_{NE}$. Choosing a value of $\alpha_{NE} = 0.04$ corresponds to a stimulation with a frequency of 2 Hz and yields a release of norepinephrine like the one measured by Mokrane et al. [18]. Figure 11 shows the simulated increase of [NE] and the corresponding simulated hr response. Any increase in heart rate yields
an increase in cardiac output $Q$ (see equation (6)) and thus an increase in blood pressure $P_a$ (see equation (2)). The impact of changes in heart rate on these cardiac parameters are simulated and the results are shown in figure 12. The mean value of cardiac output $Q$ is used for simulations to show the increase in this variable and in blood pressure $P_a$ more clearly.

5. Validation

For the validation of the model, experimental data of [24–29] were used as reference examples and compared to numerical simulations. These experimental results reflect special physiological situations and were not used to calibrate the model parameters.

The numerically simulated CO$_2$ reactivities for different values of mean arterial blood pressure and intracranial pressure are shown in figure 13. The top left of figure 13 shows the dependence of CBF under intracranial normotension ($P_i = 4$ mm Hg) and different values of MAP: $A_1$—$P_a = 97$ mm Hg, $A_2$—$P_a = 77$ mm Hg, and $A_3$—$P_a = 64$ mm Hg. Decreasing the mean arterial blood pressure provides a decreased CBF for increased values of $P_{aCO_2}$. Hauerberg et al. [24] measured the dependence of cerebral blood flow on $P_{aCO_2}$ for the above given intracranial and blood pressure values. They also got a qualitative decreased CBF for decreased arterial and increased CO$_2$ pressure values.

The top right of figure 13 shows the dependence of cerebral blood flow on arterial CO$_2$ pressure under intracranial hypertension ($P_i = 31$ mm Hg) and $B_1$—$P_a = 104$ mm Hg, $B_2$—$P_a = 123$ mm Hg. Increasing the mean arterial blood pressure yields an increase in CBF for higher values of $P_{aCO_2}$. This change in CBF was also measured by Hauerberg et al. [24] and they stated that the groups $A_1$ and $B_2$ are similar.

The bottom of figure 13 shows the CO$_2$ reactivity for a further increased intracranial pressure of 50 mm Hg and $C_1$—$P_a = 104$ mm Hg and $C_2$—$P_a = 123$ mm Hg. Increasing the mean arterial blood pressure yields an increase in CBF for higher values of $P_{aCO_2}$, as can be seen in Hauerberg et al. [24], who also stated the similarity of the groups $A_2$, $B_1$ and $C_2$.

In figure 14 a simulated autoregulation curve is shown. For validation of the lower autoregulation limit, which lies in the range of 40 and 55 mm Hg, a work of Waschke et al. [25] is used. Corresponding to their experimental studies the mean arterial blood pressure is decreased from 116 to 86, 70, 55 and 40 mm Hg and the mean cerebral blood flow is calculated for the given pressure values by the model. In figure 15 the experimental results of Waschk et al. [25] are shown and compared with the simulated data. A significant decrease in cerebral blood flow, if blood pressure is decreased from 55 to 40 mm Hg, can be seen in the measured data as well as in the simulated data.

For validation of the upper autoregulation limit, which lies in the range of 150 and 160 mm Hg, a work of Schaller et al. [26] is used. The data of their control group of the
investigated Wistar rats is compared with simulation results in figure 16, where mean arterial pressure is increased up to 10, 20, 30, 40 and 50 percent of its basal value and the corresponding cerebral blood flow is measured and calculated, respectively. One can see the increase in cerebral blood flow if blood pressure is increased up to 40–50% of its basal value in the measured data as well as in the simulated data.

Elevated intracranial pressure, the most serious acute sequela of traumatic brain injury, has direct impact on the lower autoregulation limit. Corresponding to the measured data of Engelborghs et al. [27] the intracranial pressure $P_{ic}$ was changed from 6 to 33 mm Hg in the first 5 min after head injury. After another 23 h 55 min $P_{ic}$ was decreased to 28 mm Hg. The changes in arterial CO2 pressure and in heart rate $b$ in the first day after brain damage, were

![Figure 12](image12.png)

Figure 12. Left: simulated increase in blood pressure $P_a$. Right: simulated cardiac output response, to a stimulation interval of 0–200 s and with a stimulation intensity of $\alpha_{SN}=0.04$.

![Figure 13](image13.png)

Figure 13. Top left: simulated dependence of CBF on $P_{aCO2}$ for $P_{ic}=4$ mm Hg and $A_1—P_a=97$ mm Hg, solid line, $A_2—P_a=77$ mm Hg, dashed line and $A_3—P_a=64$ mm Hg, dotted line. Decreasing MAP under intracranial normotension results in a decrease in CBF for increased values of $P_{aCO2}$ (cf. [24]). Top right: simulated dependence of CBF on $P_{aCO2}$ for $P_{ic}=31$ mm Hg and $B_1—P_a=104$ mm Hg, dashed line, $B_2—P_a=123$ mm Hg, solid line. Increasing MAP yields an increase in CBF for higher values of $P_{aCO2}$. The simulation results of $B_2$ and $A_1$ are similar. These phenomenons were also measured by Hauerberg et al. [24]. Bottom: simulated CO2 reactivity for $P_{ic}=50$ mm Hg and $C_1—P_a=104$ mm Hg, dotted line, and $C_2: P_a=123$ mm Hg, dashed line. Increasing MAP yields an increase in CBF for higher values of $P_{aCO2}$. The simulation results of $C_2$, $A_2$ and $B_1$ are similar. These phenomenons were also measured by Hauerberg et al. [24].
simulated with the given data of Waschke et al. [25]. The mean arterial blood pressure \( P_a \) was decreased linearly to 86, 70, 55 and 40 mm Hg and the cerebral blood flow under these conditions was calculated after another period of 1 h 25 min.

Figure 17 shows the dependence of cerebral blood flow on mean arterial blood pressure in the situation of brain damage described above. The calculated lower autoregulation limit increased from 40, 55 mm Hg under basal conditions to 70, 86 mm Hg. Engelborghs et al. [28] measured a lower autoregulation limit of 46.9 ± 12.7 in sham rats and 62.2 ± 20.8 in CHI rats.

Figure 18 shows the effect of L–NAME infusion on blood flow in the rat cerebral cortex. Hudetz et al. [29] measured the laser–Doppler flow (LDF) thirty minutes after N\(^\text{\textit{n}}\)–nitro-L-arginine methyl ester (L-NAME) administration (20 mg/kg). LDF was reduced from 159 ± 14 to 135 ± 11 perfusion units (PU) (15% decrease), whereas mean arterial pressure \( P_a \) was increased from 105 ± 4 to 132 ± 6 mm Hg (26% increase).

This effect of nitric oxide inhibition is numerically simulated in the way that the following parameters are changed linearly in thirty minutes: the basal value of blood pressure \( P_a = 116 \) mm Hg is increased to 146.16 mm Hg (26% increase) and the basal NO production rate \( q_{\text{NO}} = 54.1 \) ng/g tissue is decreased to 5.4 ng/g tissue (90% decrease). The cerebral blood flow was calculated after another period of thirty minutes and it was decreased from 0.1696 to 0.1393 ml s\(^{-1}\) (17.8% decrease).

6. Discussion

In this paper, a mathematical model is presented which describes cerebral hemodynamics under physiological aspects. The overall aim is to develop a tool that provides a more complete insight into the mechanisms of cerebrovascular perfusion with special emphasis of the interaction of different regulatory mechanisms. To facilitate hypothesis testing in the lab a species specific model using data from the Sprague-Dawley rat was developed. In the first instance, only two of the most robust mechanisms have been integrated into the model: CO\(_2\)- and NO-regulation.

By introducing a cardiac output function as input for the systemic circulation the blood pressure gets dependent on time and thus all pressures of the model and the blood flow become pulsatile. The implementation of pulsatile blood flow allows to investigate the effects of cardiac dysrhythmias on cerebral perfusion which is an important problem with respect to the increasing population of elderly trauma victims with cardiac comorbidities. The systemic circulation is closed for the first time in this model, so all changes in cardiac output result in changes of other systemic pressures and flows.

Maintenance of an adequate cerebral perfusion pressure is a mainstay of TBI (traumatic brain injury)
therapy and is thought to be a necessary and potentially neuroprotective therapeutic intervention that “may be associated with a substantial reduction in mortality and improvement in quality of survival and is likely to enhance perfusion to ischemic regions of the brain after severe TBI” (cf. [30]). This goal can be reached by intravenous application of vasopressors (e.g. norepinephrine). The immediate effector organ of vasopressors is the heart where an increase of cardiac output results from modulation of heart rate and contractility. Interestingly there appears to be a feedback from TBI to cardio-circulatory performance in terms of a reduced response to vasopressors under certain conditions (cf. [22]). Thus, the system of sympathetic regulation of heart function needs to be implemented into a physiologically relevant model of cerebral perfusion. In the present version of the model only the very basic mechanism of norepinephrine physiology was considered as the model is focused only on the most basic and robust mechanisms of cerebral perfusion (flow, autoregulation). After the fundamental behaviour of cerebral blood flow has been successfully modelled future versions of the model need to map the effects of vasopressors more precisely.

Other important features of the model are the regulation mechanisms, like autoregulation, \(CO_2\) reactivity and NO reactivity, which regulate the cerebral blood flow under changes in arterial \(CO_2\) pressure, endothelial NO production and changes in CBF itself.

Several considerations are made to extend the model in the following way: the point of extension deals with the assumptions which are made for NO reactivity. It is important to note that NO is also produced at other regions in brain and thus has an effect on the smooth muscle cells of, i.e. large cerebral arteries and veins, and yields a change in compliance and resistance of these vessels. There is also a change in stroke volume \(\nu\), heart rate \(b\) and systemic resistance \(R\), as a consequence of NO production at the vessel cells of central veins (i.e. vena cava), which has to be taken into account. However, NO is treated as an input quantity of the system. This means that the distinct sources of NO generation have not been modelled explicitly. A thorough model of NO generation needs to comprise a variety of cell types (endothelial cells, astrocytes, etc.) as well as a complicated system of subcellular structures. Thus the simplification of NO physiology, i.e. to consider NO “to be simply there”, appears justified. A multiscale submodel of NO physiology would complicate the model as a whole substantially. A multiscale submodel would only be justified in terms of additional information if a spatial resolution of the cerebrovascular system were given as these subcellular events are governed by the local milieu. In the present model, variation of NO concentrations within physiologic bounds will provide the necessary information.

Figure 16. Left: measured data of Schaller et al. [26]: The mean arterial blood pressure \(P_a\) is raised up to 10, 20, 30, 40 and 50 percent of its basal value (116 mm Hg) and the mean regional cerebral blood flow is measured for the corresponding pressure values in LD-Units. Right: calculated mean cerebral blood flow for the given blood pressure values by the model. The increase in cerebral blood flow if blood pressure is increased up to 40 and 50% of its basal value is qualitatively the same and therefore the upper regulation limit must be in the range of 150 and 160 mm Hg.

Figure 17. Numerically simulated dependence of CBF on mean arterial pressure under the situation of a head injury. The values of the intracranial pressure, \(P_{\text{ICP}}\), and \(b\) under this condition are given by [27] and [25], respectively. The lower autoregulation limit increased and lies in the range of 70 < \(x\) < 86 mm Hg.
References


