Modelling and Simulation of the Human Cardiovascular System 
by Differential Hybrid Petri Net 

André A. Jorge*, Marcosiris A. O. Pessoa*, Fabrício Junqueira*, Luis A. M. Riascos**, 
Diolino J. Santos Filho*, Paulo E. Miyagi*

*Escola Politecnica da Universidade de Sao Paulo, Brazil 
(e-mails: andre.anjorge@gmail.com, marcosiris@usp.br, fabri@usp.br, diolinos@usp.br, pemiyagi@usp.br) 
**Universidade Federal do ABC, Brazil 
(e-mail: luis.riascos@ufabc.edu.br)

Abstract: Clinical data surveys indicate a significant number of deaths deriving from diseases in the human cardiovascular system (HCS). This is one of the main motivations for identifying problems in this system and ways for solving them totally or partially. In this work, HCS is modeled as a hybrid system (discrete event systems combined with systems of continuous variables) for a more detailed characterization of this system functioning, by applying the formalism of the Differential Hybrid Petri Net (DHPN). The analysis of this model, using Matlab®/Simulink, presented consistent results when compared with clinical data, regarding physiological variables, such as blood pressure and blood flow rate, indicating the validity of this model.

Keywords: Human cardiovascular system, Hybrid systems, Differential hybrid Petri nets, Lumped parameter model, Blood pressure, Blood flow rate.

Nomenclature

<table>
<thead>
<tr>
<th>Letter</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>C</td>
<td>Capacitance</td>
</tr>
<tr>
<td>E</td>
<td>Complacency of the vessel wall</td>
</tr>
<tr>
<td>D</td>
<td>Diode (Valve)</td>
</tr>
<tr>
<td>E</td>
<td>Elastance</td>
</tr>
<tr>
<td>HR</td>
<td>Heart Rate</td>
</tr>
<tr>
<td>I</td>
<td>Electrical current</td>
</tr>
<tr>
<td>K</td>
<td>Coefficient</td>
</tr>
<tr>
<td>L</td>
<td>Inductance</td>
</tr>
<tr>
<td>C</td>
<td>Blood inertia</td>
</tr>
<tr>
<td>P</td>
<td>Blood pressure</td>
</tr>
<tr>
<td>Q</td>
<td>Blood flow rate</td>
</tr>
<tr>
<td>q</td>
<td>Charge</td>
</tr>
<tr>
<td>R</td>
<td>Electrical resistance</td>
</tr>
<tr>
<td>R</td>
<td>Viscous flow resistance</td>
</tr>
<tr>
<td>T</td>
<td>Time</td>
</tr>
<tr>
<td>V</td>
<td>Volume</td>
</tr>
<tr>
<td>v</td>
<td>Voltage</td>
</tr>
</tbody>
</table>

Subscripts

<table>
<thead>
<tr>
<th>Symbol</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>Initial Value; Offset Value</td>
</tr>
</tbody>
</table>

1. INTRODUCTION

According to the World Health Organization (2017), cardiovascular diseases (CVDs) are still the number one cause of death globally; more people die annually from CVDs than from any other cause. Thus, many studies for identifying problems in the human cardiovascular system (HCS) and ways for solving them, totally or partially, have increased over the years (Jorge et al., 2018, 2019a, 2019b).

The laws of blood fluid dynamics, called hemodynamic, govern the cardiovascular system. Hemodynamic may be described by three parameters: blood flow (cardiac output), blood pressure and vascular resistance. The measurement of these hemodynamic parameters is the foundation for diagnosing different CVDs (Westerhof et al., 2005).

Since there is a growing demand for restrictions on in vivo testing for the medical equipment industry, including the case of animal testing (Watanabe et al., 2014), new strategies...
and platforms for in vitro tests and tools for evaluating implantable medical devices, such as ventricular assist devices (VADs), must be considered (Jorge et al., 2018, 2019a, 2019b).

Knowing that any closed fluid system has an analogy with electrical circuits (Westerhof et al., 2005), the cardiovascular system may be considered analogous to electrical circuits. Thus, lumped parameter models, known as 0D model or electrical circuits, of the HCS are an appropriate method to obtain several hemodynamic parameters of the blood circulation (Formaggia et al., 2009; Rahman and Haque, 2012; Gul, 2016).

Hybrid systems are defined as systems in which state variables of a continuous and discrete nature are simultaneously found. The evolution of the system may occur partially as a function of time and/or as a function of the discrete event occurrence. This means that, in a hybrid system, subsystems belonging to the two classes presented coexist simultaneously (Villani et al., 2007).

In general, the approaches to the modeling of hybrid systems consist in extensions of continuous models, such as ordinary differential equations, in which variables whose value may be modified in a discontinuous way in time, are included. Other approaches consist in modifying modeling techniques applied to discrete event systems, whereby new elements are introduced to represent the continuous dynamics of the system, such as in the models based on Hybrid Petri net (David and Alla, 2010). In addition, there are also intermediate approaches that combine models of continuous systems, described by differential equations, and discrete systems, described by finite automata or Petri net (Giua and Silva, 2017; Boussif and Ghazel, 2018), in which an interface is established for the communication between the two types of models (Jorge et al., 2019a, 2019b). However, these works (Jorge et al., 2019a, 2019b) have a limitation to modeling cardiovascular diseases, considered as discrete events, due to the modelling approach used (0D model).

This work proposes a model for the HCS applying the formalism of the Differential Hybrid Petri net (DHPN), using the equations derived from the lumped parameter model previously developed and presented in (Jorge et al., 2019a, 2019b). The goal of this contribution is proposing a valid model of the HCS to study the behaviour of this system, taking into account the interactions among their dynamics (continuous system combined with discrete events). This proposed model was simulated and analysed using Matlab®/Simulink. The focus lies on the analysis and comparison of the blood flow rate and blood pressure from clinical data with simulated data.

### 2. METHODOLOGY

In Jorge et al. (2019a, 2019b), the lumped parameter model or the 0D model was applied to model the HCS, since this technique represents physiological variables, such as blood pressure, blood flow rate, viscous flow resistance and others, of the blood circulation system, presenting an immediate solution.

According to Formaggia, Quarteroni and Veneziani (2009) and Rahman and Haque (2012), each segment of a blood vessel may be represented by a resistor-inductor-capacitor (RLC) electrical circuits (0D model). Table 1 summarizes the relationship among the variables considered (Gul, 2016; Malatos, Raptis and Xenos, 2016; Jorge et al., 2019a, 2019b).

<table>
<thead>
<tr>
<th>Fluid Dynamics</th>
<th>Physiological Variables</th>
<th>Electrical System</th>
</tr>
</thead>
<tbody>
<tr>
<td>Pressure (P)</td>
<td>Blood pressure (P)</td>
<td>Voltage (V)</td>
</tr>
<tr>
<td>Flow rate</td>
<td>Blood flow rate (Q)</td>
<td>Current (I)</td>
</tr>
<tr>
<td>Volume</td>
<td>Blood volume (V)</td>
<td>Charge (q)</td>
</tr>
<tr>
<td>Viscosity</td>
<td>Viscous flow resistance (R)</td>
<td>Resistance (R)</td>
</tr>
<tr>
<td>Inertance</td>
<td>Blood inertia (L)</td>
<td>Inductance (L)</td>
</tr>
<tr>
<td>Elastic coefficient</td>
<td>Complacency (c)</td>
<td>Capacitance (C)</td>
</tr>
</tbody>
</table>

This model considers the blood leaves the left ventricle, flows through the systemic circulation (body except lungs) into the right part of the heart (atrium and ventricle) and from there through the vessels of the pulmonary circulation (lungs) back into the left atrium and ventricle (Guyton and Hall, 2011) and (Tortora and Derrickson, 2014).

The parameter values for the proposed lumped parameter model (Fig. 1) are derived from (Korakianitis and Shi, 2006a, 2006b) and adjusted for this model. However, several difficulties make parameter setting a challenging task, such as: the invasive nature of many of the measurements, restricted access to the required measurement sites due to anatomical configuration, practical difficulties in the orientation of flow probes (particularly invasive ones), difficulties in synchronizing pressure and flow data (particularly when they are not measured simultaneously), limited precision in the pressure/flow sensors, all of which contribute to the accuracy of the model parameters (Shi, Lawford and Hose, 2011). Perhaps more important, the data available of the pressure measurement provides only part of the information needed to estimate the model parameters. Thus, it is necessary to develop more accurate and efficient techniques to optimize the parameter setting in lumped parameter models (Jorge et al., 2019a, 2019b).

#### 2.1 Human Cardiovascular System

The model of pulsatile human blood circulation consists of a pumping heart coupled to lumped descriptions of the systemic and the pulmonary circulation (loop). The ventricles are guided by a pair of time-varying elastance functions, whereas the two atria are purely passive chambers. In addition, there are four heart valves (mitral, aortic, tricuspid and pulmonary) in these chambers, as shown in Fig. 1. The valves allow a small amount of volume to flow back into the left and right atria and ventricles before closure is completed. When the left atrial pressure exceeds the left ventricular pressure, the mitral valve (MV) opens and the left ventricle is filled with blood. After that, when left ventricular pressure exceeds the root aortic pressure, the aortic valve (AV) opens and blood flows through the systemic circulation, consisting of aorta (A), systemic arteries (S4), systemic arterioles (SAR), systemic capillaries (SC), systemic veins (SV) and vena cava (VC). The veins return the blood to the passive right atria. Next, when the right atrial pressure exceeds the...
right ventricular pressure, the tricuspid valve (TV) opens and the right ventricle is filled with blood. Then, when the right ventricular pressure exceeds the root pulmonary artery pressure, the pulmonary valve (PV) opens and blood flows through the pulmonary circulation, consisting of pulmonary artery (PA), pulmonary arteries (PAS), pulmonary arterioles (PAR), pulmonary capillaries (PC), pulmonary veins (PVS) and pulmonary vein (PAV), returning to the left atrium and starting a new cycle.

2.1.1 Heart

The cardiac contractile properties of the atria and ventricles are assumed to be defined by time-varying elastance functions. The relation between left ventricular cavity pressure ($P_{LV}$) and ventricular volume ($V_{LV}$) is described by (Ferreira et al., 2005):

$$P_{LV} = E(t) (V_{LV} - V_{LV,0}),$$

(1)

where $V_{LV,0}$ represents the left ventricular volume at zero pressure. The elastance function $E(t)$ is given by (Ferreira et al., 2005):

$$E(t) = (E_{max} - E_{min}) \left(1.55 \left(1 + \frac{t_c}{T_{max}}\right)^{1.9} + \frac{1}{1 + \frac{t_c}{T_{max}}^{1.9}}\right) + E_{min},$$

(2)

In (2), $t_c = t/T_{max}$, $T_{max} = 0.2 + 0.15t_c$, and $t_c$ is the cardiac cycle interval, i.e., $t_c = 60/HR$, where $HR$ is the heart-rate. Notice that constants $E_{max}$ and $E_{min}$ are related to the end-systolic pressure volume relationship (ESPVR) and the end-diastolic pressure volume relationship (EDPVR), respectively (Ferreira et al., 2005).

The flow rate in the aortic valve is described by:

$$\frac{dQ_{AV}}{dt} = \frac{P_{LV} - P_{a} - (R_{LV} + R_{av})Q_{AV}}{L_{LV}}, \quad \text{aortic valve is open}$$

$$Q_{AV} = 0, \quad \text{aortic valve is closed},$$

(3)

The behaviour of the left and right atria, the right ventricle and the other valves are modeled by a similar description.

2.1.2 Notation for the Cardiovascular Model

The notation used in the electrical analogy of the cardiovascular system is the same adopted by Formaggia, Quarteroni and Veneziani (2009). From Fig. 2, the governing equations associated to this generic single segment of a blood vessel are given by (Blanco and Feijóo, 2010):

$$\frac{dQ_{i}}{dt} + R_{i}Q_{i} = P_{i} - P_{o},$$

(4)

$$C \frac{d}{dt}(P_{i} - P_{ex}) = Q_{i} - Q_{o},$$

(5)

where $P_{i}$ and $P_{o}$ are the input and output pressures, $P_{ex}$ is the external pressure, $Q_{i}$ and $Q_{o}$ are the blood inflow and outflow.

2.2 Differential Hybrid Petri Nets

A DHPN is composed of two types of places and two types of transitions: discrete places and transitions,
differential places and transitions, see Fig. 3. A DHPN has three types of arcs: normal arc, test arc and inhibitor arc. The normal arc and the inhibitor arc can be used to connect discrete elements, and the test arc is used for connecting differential places and discrete transitions.

Fig. 3. Places and transitions of DHPN.

A DHPN is defined by a 14 tuple \( (P_0, T_D, P_D, T_{DF}, X, A_S, A_I, A_T, P_{NS}, P_{on}, T, H, J, M_0, \) where (Sousa and Lima, 2008; Yu, Dou and Li, 2016):

\[
P_0 = \{P_1, P_2, ..., P_{n}\} \text{ is a finite and non-empty set of discrete places, which represents the operation modes of all units;}
\]

\[
T_D = \{T_1, T_2, ..., T_{n}\} \text{ is a finite and non-empty set of discrete transitions, which represents the event-triggered switching behaviours;}
\]

\[
P_{DF} = \{P_{1f}, P_{2f}, ..., P_{nf}\} \text{ is a finite and non-empty set of differential places, which describes the continuous states of all units;}
\]

\[
T_{DF} = \{T_{1f}, T_{2f}, ..., T_{nf}\} \text{ is a finite and non-empty set of differential transitions, which represents the dynamic behaviours;}
\]

\[
P = P_D \cup P_{DF}, T = T_D \cup T_{DF}, P \cap T = \emptyset \text{ or } P \cup T \neq \emptyset,
\]

\[
X = [x_1, x_2, ..., x_n]^T, \text{ is a continuous state vector of } P_{DF};
\]

\[
A_S \subseteq ((P_D \times T_D) \cup (T_D \times P_D)) \cup ((P_D \times T_{DF}) \cup (T_{DF} \times P_D)) \text{ is a set of normal arcs;}
\]

\[
A_I \subseteq (P_D \times T_D) \text{ is a set of inhibitor arcs;}
\]

\[
A_T \subseteq (P_D \times T_D) \cup (T_D \times P_{DF}) \text{ is a set of test arcs;}
\]

\[
Pre (P_s, T) P_D \times T_D \rightarrow N \text{ is called the predecessor function, which defines the normal arcs of a place to a transition;}
\]

\[
Pos (P_s, T) P_D \times T_D \rightarrow N \text{ is called a successor function, which defines the normal arcs of a place to a transition;}
\]

\[
T = \text{ a timing map for the discrete transitions, by which the minimal switching interval can be defined;}
\]

\[
H_{P,T} : P_{DF} \times T_D \rightarrow X \text{ is a junction function associated with the test arc, which connects a differential input place } P_i \text{ to a discrete transition } T_j;
\]

\[
J_{T_i,P_0} : T_D \times P_{DF} \rightarrow X \text{ is an enabling function associated with the test arc, which connects a transition } (T_i \in T_D) \text{ to a differential output place } (P_0 \in P_{DF});
\]

\[
M_0(t) : P \rightarrow N, \text{ represents the initial marking of the net in } t = 0.
\]

A model proposal for the HCS, by applying the formalism of the DHPN, was developed, as presented in Fig. 4. In this model, the enabling functions \( H_{P,T} : P_{LA} \geq P_{LV} \) and \( H_{P,T} : P_{LA} < P_{LV} \) enable opening and closing the Mitral Valve, depending on when the left atrium pressure is greater or equal or less than the left ventricle pressure, respectively. From the differential equations of the 0D model, derived by manipulation of (4) and (5), the differential continuous state vector of \( T_{DF} \) and the continuous state vector of \( P_{DF} \) are obtained, as indicated in Fig. 4. The other enabling functions for the other valves and the other differential continuous state vectors have a similar approach.

Fig. 4. Modelling of the human cardiovascular system by Differential Hybrid Petri Net.
3. RESULTS AND DISCUSSION

In order to evaluate the capability of the proposed model to emulate the HCS hemodynamics, tests are performed by simulation using Matlab®/Simulink. These numerical simulations ignore the effects of gravity; one may hence assume that blood flows solely in response to pressure gradients. In addition, the four heart valves were simulated with the traditional on–off valve model. Thus, the normal reverse flow in the heart valves, following valve closure, is not presented in the results. In addition, normal functioning and abnormal situations, such as the aortic blockage generated by atherosclerosis or thrombosis, were simulated with the proposed model. However, due to lack of space, only the results of the normal functioning of the HCS are presented here.

Figure 5 shows the left cardiac cycle from clinical data (Guyton and Hall, 2011).

Figure 6 shows simulated hemodynamic waveforms throughout the HCS model. The comparison of the numerical simulation for the left cardiac cycle (Fig. 6(a)) with clinical data (Fig. 5) shows good agreement. Moreover, the results for the right cardiac cycle and flow rate into the heart and the systemic and pulmonary circulation are in agreement with corresponding human data found in the literature (Noordergraaf, 1978; Korakianitis and Shi, 2006a, 2006b; Ferreira et al., 2005; Blanco and Feijoo, 2010; Nichols, O’Rourke and Vlachopoulos, 2011; Guyton and Hall, 2011; Friselli et al., 2015; Mynard and Smolich, 2015; Bakir et al., 2018).

Therefore, the results are consistent with clinical data and related papers, and indicate the validity of the proposed model for the HCS by using the hybrid system approach and DHPN as a modelling technique.

4. CONCLUSIONS

This work proposed a model for the human cardiovascular system (HCS), by applying the formalism of the Differential Hybrid Petri Net (DHPN), since this formalism takes into account the concept of hybrid systems for a more detailed characterization of the cardiovascular system functioning. This model was implemented in Matlab®/Simulink.

The numerical simulations show simulated hemodynamic waveforms throughout the cardiovascular system model. The results are consistent with clinical data and related papers, indicating the validity of the proposed model for the human cardiovascular system. Thus, the application of the formalism DHPN to model the HCS allows a better analysis of the behaviour of this system, when taking into account the combination between continuous and discrete parts.

Future works will consider replacing the traditional on–off valve model by dynamic valves and including events, such as variations in the body positions and the cardiovascular diseases, to analyse the simulated hemodynamic throughout in the HCS model compared with clinical data and related papers.
ACKNOWLEDGEMENT

The authors gratefully acknowledge the financial support provided by the Coordenação de Aperfeiçoamento de Pessoal de Nível Superior - Brasil (CAPES) - finance code 001, grant #2013/24434-0, Sao Paulo Research Foundation (FAPESP) and Brazilian National Council of Scientific and Technological Development (CNPq).

REFERENCES


