Supplementary materials 1

Supplementary methods

1 Heart model

1.1 Mesh generation:

Segmentations of the multi-detector CT scan data taken at end-diastole were performed using commercial software (VirtualPlace; AZE Ltd., Tokyo, Japan). Based on these data, finite element meshes of the ventricles and torso were created. We used voxel meshes of the heart (size 0.2mm) and torso (size 1.6m) for electrophysiology and a tetrahedral mesh of the ventricles (size 2mm) for mechanical analyses. To these heart models, fiber orientations were mapped using the rule-based method[1].

1.2 Electrophysiology:

We solved the following bidomain equations

$$
\beta \left(C_m \frac{\partial V}{\partial t} + I_{ion} \right) = I_{stim} - \frac{\partial}{\partial x_i} \left(G_{ij}^E \frac{\partial \phi^E}{\partial x_i} \right) \tag{1}
$$

$$
\beta \left(C_m \frac{\partial v}{\partial t} + I_{ion} \right) = -I_{stim} - \frac{\partial}{\partial x_i} \left(G_{ij}^I \frac{\partial \phi^I}{\partial x_i} \right) \tag{2}
$$

where Φ^E and Φ^I are the extra- and intra cellular potential respectively, $V = \Phi^I - \Phi^E$ is transmembrane voltage, β is the surface-to-volume ratio of the tissue, Cm is membrane capacitance, t is time, G_{ij}^E and G_{ij}^I are the intra- and extracellular anisotropic conductivity tensors, x_i and x_i are the tensor notation of the x, y, and z coordinates, and I_{ion} is the sum of ionic transmembrane currents calculated by the human ventricular myocyte model of electrophysiology. Values of conductivity used for the heart and torso were adopted from literature [2-4] (supplementary Table 1). We used the human cell model of electrophysiology by tenTusscher et al. [5] for ventricular tissue. The Purkinje network was modeled as a thin layer with high conduction velocity on the endocardial surface [6-8].

1.3 Sarcomere model:

We used the spatially-detailed model of sarcomere dynamics we developed [9-11], in which 38 myosin heads on the thick filament and 32 tropoin/tropomyosin regulatory units on the thin filament in half sarcomere were modeled. Frank-Starling mechanism was realized by the overlap of thick and thin filaments. Co-operativity among the neighboring myosin heads was introduced for the transition rates between detached and attached states in the crossbridge cycling.

1.4 Mechanics:

The discretized form of equation of motion was solved in the tetrahedral mesh:

$$
\nabla \cdot \boldsymbol{\sigma} + \boldsymbol{f} = 0 \tag{3}
$$

where σ is the Cauchy stress tensor and f is the body force. σ can be described as:

$$
\sigma = J^{-1} \mathbf{F} \mathbf{S} \mathbf{F}^T,\tag{4}
$$

where $\mathbf{F} = \frac{\partial x}{\partial x}$ is the deformation gradient tensor, *S* is the second Piola-Kirchoff stress and *J* $=$ det \vec{F} is the Jacobian determinant.

S was given as:

$$
S = 2\frac{\partial W_{pass}}{\partial C} - pC^{-1} + S^a,\tag{5}
$$

where C is right Cauchy-Green deformation tensor and p is the hydrostatic pressure.

Strain energy function (Wpass) for the passive property was calculated as the sum of the two functions proposed by Humphrey et al. (W_{Hum}) [12] and Lin et al. (W_{Lin}) . [13]

$$
W_{pass} = \mathbf{a} \cdot W_{Hum} + b \cdot W_{Lin} \tag{6}
$$

$$
W_{Hum} = c_1(\alpha - 1)^2 + c_2(\alpha - 1)^3 + c_3(I_1 - 1) + c_4(I_1 - 1)(\alpha - 1) + c_5(\alpha - 1)^2
$$
 (7)

$$
W_{Lin} = c_6(e^Q - 1) \tag{8}
$$

$$
Q = c_7(I_1 - 3)^2 + c_8(I_1 - 3)(I_4 - 1) + c_9(I_4 - 1)^2
$$
\n(9)

 $I_1 = trC$, I_4 =*NCN*, $\alpha = \sqrt{NCN}$, *C* is the right Cauchy-Green deformation tensor and *N* is the unit vector defining the preferred direction of muscle in the undeformed state. Parameter values of c₁=3080pa, c₂=3240pa, c₃=359pa, c₄=-1940pa, c₅=1660pa, c₆=82.3pa, c₇=7.3, c₈=1.86 and $c_9=0.0640$ were used, and a and b were adjusted to reproduce the diastolic pressure-volume relation estimated by the formula by Klotz et al. [14] in patients.

The active stress tensor S^a was calculated as:

$$
S_a = \begin{bmatrix} S_{ff}^a & 0 & 0 \\ 0 & S_{ss}^a & 0 \\ 0 & 0 & S_{nn}^a \end{bmatrix},
$$
 (10)

where *f*, *s*, and *n* indicate fiber direction, sheet direction, and sheet normal direction, respectively. $S_{ss}^a=0$, $S_{nn}^a=0$ and S_{ff}^a was the active force calculated by the above sarcomere model.

2. Circulatory model:

The finite element model of ventricles was connected to a lumped parameter model of systemic and pulmonary circulation (Fig. 1A) similar to Kerckhoffs et al. [15] and solved by the following equations.

Systemic circulation:

$$
Pas = Vas/C1a \tag{11}
$$

$$
Pvs=Vvs/C2a \tag{12}
$$

$$
Q_{ao} = \begin{cases} \frac{P_{LV} - P_{as}}{R_{1a}} & (P_{LV} > P_{as})\\ 0 & (P_{LV} < P_{as}) \end{cases}
$$
(13)

$$
Q_{as} = (Pas - Pvs)/R2a \tag{14}
$$

$$
Q_{vs} = (Pvs - P_{RA})/R3a \tag{15}
$$

$$
Q_{\text{mitral}} = \begin{cases} \frac{P_{LA} - P_{LV}}{R4a} & (P_{LA} > P_{LV})\\ 0 & (P_{LA} < P_{LV}) \end{cases}
$$
(16)

$$
\frac{dV_{LA}}{dt} = Q_{VP} - Q_{\text{mitral}}
$$
 (17)

$$
\frac{dV_{LV}}{dt} = Q_{\text{mitral}} - Q_{\text{ao}}
$$
 (18)

$$
\frac{dV_{\rm as}}{dt} = Q_{\rm ao} - Q_{\rm as} \tag{17}
$$

$$
\frac{dV_{vs}}{dt} = Q_{as} - Q_{vs} \tag{18}
$$

Pas is the pressure of the systemic arteries, Pvs is pressure of the systemic veins, PLV is pressure in the left ventricle, PLA is pressure in the left atrium, Vas is volume of the systemic arteries, Vvs is volume of the systemic veins, VLV is volume of the left ventricle, VLA is volume of the left atrium, Q_{mitral} is mitral flow, Q_{VP} is pulmonary venous flow, Qao is aortic flow, Qas is flow out of the systemic arteries, and Qvs is flow out of the systemic veins.

Pulmonary circulation:

$$
Pap = \text{Vap}/\text{C1p} \tag{19}
$$

$$
Pvp = Vvp/C2p \tag{20}
$$

$$
Q_{pa} = \begin{cases} \frac{P_{RV} - P_{ap}}{R1p} & (P_{RV} > P_{ap})\\ 0 & (P_{RV} < P_{ap}) \end{cases}
$$
 (21)

$$
Q_{ap} = (Pap - Pvp)/R2p
$$
 (22)

$$
Q_{vp} = (Pvp - P_{LA})/R3p
$$
\n(23)

$$
Q_{\text{tricus}} = \begin{cases} \frac{P_{\text{RA}} - P_{\text{RV}}}{R_{\text{4a}}} & (P_{\text{RA}} > P_{\text{RV}}) \\ 0 & (P_{\text{RA}} < P_{\text{RV}}) \end{cases} \tag{24}
$$

$$
\frac{dV_{RA}}{dt} = Q_{vs} - Q_{tricus}
$$
 (25)

$$
\frac{dV_{RV}}{dt} = Q_{tricus} - Q_{pa}
$$
 (26)

$$
\frac{dV_{ap}}{dt} = Q_{pa} - Q_{ap} \tag{27}
$$

$$
\frac{dV_{vp}}{dt} = Q_{ap} - Q_{vp} \tag{28}
$$

Pap is the pressure of the pulmonary arteries, Pvp is pressure of the pulmonary veins, P_{RV} is pressure in the right ventricle, P_{RA} is pressure in the right atrium, Vap is volume of the pulmonary arteries, Vvp is volume of the pulmonary veins, V_{RV} is volume of the right ventricle, V_{RA} is volume of the right atrium, Q_{tricus} is tricuspid flow, Q_{vs} is systemic venous flow, Q_{pa} is pulmonary artery flow, Qap is flow out of the pulmonary arteries, and Qvp is flow out of the pulmonary veins. The right and left atria were modeled by the time-varying elastance model proposed by Kaye et al. [16].

3. Parameter personalization

3.1 ECG personalization:

We adjusted parameters to make a match between the simulated ECG and the actually recorded standard 12-lead ECGs of patients following the previously reported procedure [7]. First, we fitted the QRS complex by shifting standard endocardial earliest activation sites determined in our previous study[7]. In the case of left bundle branch block, endocardial earliest activation sites located in the right side of the heart were used. Next, the transmural distribution of different cell types, namely endocardial, M-, and epicardial cells, were adjusted to reproduce T-wave morphologies. Goodness of the fit was evaluated by visual inspection and the calculation of

cross-correlation (Rcc) between the simulated and actual ECGs.

$$
R_{NCC} = \frac{\sum_{j=1}^{16} \sum_{i=1}^{N} A(i,j) \times B(i,j)}{\sqrt{\sum_{j=1}^{16} \sum_{i=1}^{N} A(i,j)^2 \times \sum_{l=1}^{16} \sum_{k=1}^{N} B(k,l)^2}}
$$
(29)

 $Rcc > 0.8$ was taken as the sign of good agreement.

3.2 Personalization of systemic and pulmonary circulation:

We estimated initial values of vascular parameters based on ultrasonic cardiogram (UCG) recording, arterial blood pressure measurement and literature.

Stroke volume was calculated as the product of the end-diastolic volume derived from the finite element model and the ejection fraction by UCG. Cardiac output was then calculated by multiplying the stroke volume by heart rate. Mean arterial pressure (Pm) was estimated from systolic pressure (Ps) and diastolic pressure (Pd) by the following equation

$$
Pm = \frac{(Ps - Pd)}{3} + Pd,\tag{30}
$$

and mean right atrial pressure was estimated from the diameter of inferior vena cava measured by UCG [17]. Total systemic resistance (=R1a+R2a+R3a) was calculated as (mean arterial pressure– mean right atrial pressure)/cardiac output. At this point we assigned 93% of total systemic resistance to R2a, 5% to R1a, and 2% to R3a [18-20]. Assuming that end-systolic pressure (Pes)

was equal to mean arterial pressure, we estimated the time constant of arterial pressure decay during diastole $(τ)$ by fitting to the exponential function:

$$
Pd = Pes \cdot e^{-\frac{t_d}{\tau}},\tag{31}
$$

where t_d is the diastolic time interval obtained from the UCG recording. [21] C1a was given by dividing the τ value by R2a. Guyton et al. estimated the ratio between venous and arterial capacitances at 30 in live dogs [22] while it was set at 70 in simulation of human heart failure by Kaye et al. [16]. We adopted the intermediate value of 40 in this study. Filling resistance of the tricuspid valve (R4a) was set at 0.0025mmHgs/ml [16].

Parameter values for the pulmonary circulation were determined in a similar way. 37%, 43% and 20% of the calculated total pulmonary resistance was assigned to R1p, R2p and R3p, respectively, and mitral filling resistance was set at 0.0025mmHg s/ml according to Kaye et al. [16]. Although pulmonary venous capacitance $(C2p)$ and arterial capacitance $(C3p)$ varied among studies, we adopted the value by Kaye et al. $(C2p=7, C3p=4 \text{ ml/mmHg}$ for 70kg body weight) and adjusted for the patient's weight. Stressed blood volume was set at 15% of total blood volume which was assumed to be 8% of body weight.

Similar to Kerckhoff et al. [15], we tuned the initial parameter values using a simpler model in which the finite element model of the ventricles in supplementary Fig. 1 was replaced with time-varying elastance models of the right and left ventricles [23]. Because this system runs faster than real time, we could tune the parameter values efficiently. The program was written by LabVIEW 2012 (National Instruments, Austin, TX, USA) and the tuning was performed interactively starting from initial values.

Finally, parameter values thus determined were applied to the finite element model simulation. At this stage, contractility of the heart model was adjusted by multiplying adjusting factors to the active stress tensor *S^a* .

Supplementary Table 1

Vascular parameters

Supplementary Table 3

CRBBB: complete right bundle branch block, CLBBB: complete left bundle branch block, DCM:

dilated cardiomyopathy, OMI: old myocardial infarction, DCM: dilated cardiomyopathy

NYHA class was not evaluated for the patient 4 due to the gait disturbance.

Supplementary references

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