Population modeling approach to study agerelated effects on the excitation-contraction coupling in human cardiomyocytes A.Dokuchaev, S. Khamzin, O.Solovyova

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Motivation

Identify the dominant factors in the progress of heart disease during aging



TNNP+M electromechanical model of cardiomyocyte



- Katsnelson, L. B., Solovyova, O., Balakin, A., Lookin, O., Konovalov, P., Protsenko, Y., ... Markhasin, V. S. (2011). Contribution of mechanical factors to arrhythmogenesis in calcium overloaded cardiomyocytes: Model predictions and experiments. Progress in Biophysics and Molecular Biology, 107(1), 81–89. https://doi.org/10.1016/j.pbiomolbio.2011.06.001
- ten Tusscher, K. H. W. J. (2006). Alternans and spiral breakup in a human ventricular tissue model. AJP: Heart and Circulatory Physiology, 291(3), H1088–H1100. https://doi.org/10.1152/ajpheart.00109.2006
- 3. N. Balakina-Vikulova, O. Solovyova, A. Panfilov, L. Katsnelson. Mechano-Electric Feedbacks in a New Model of the Excitation-Contraction Coupling in Human Cardiomyocytes. Computing in Cardiology, V.45, 2018.

Model population generation

- 9 parameters in range of 0-200%:
 - o gNa
 - o gCaL
 - **gK1**
 - gto
 - gKr
 - o gKs
 - SERCA
 - NaCaX
 - NaKX
- 20000 models, LHS sampling
- 200 isometric cycles for each model to reach steady-state + 1 isotonic cycle



Sampling methods

Monte Carlo sampling (MC)

- + simplicity
- + normal distribution (more "natural")
- low coverage rate $1/\sqrt{(N)}$
- normal distribution (clustering)

Latin Hypercube Sampling (LHS)

- + coverage rate ~ 1/N
- + cover full range of parameters
- + no clustering





Models filtering



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Population filtering

• AP filtering 4096 models



Population filtering

• AP+Ca filtering **1028 models**



Software

• Sampling



Computation

Latin-Hypercube(pyDOE.lhs())



SUNDIALS: SUite of Nonlinear and DIfferential/ALgebraic Equation Solvers

Numba@jitclass()

• Postprocessing and data storage







Studying the effects of aging

- $I_{to} \neq [1]$
- V_{max} SERCA ↓ [2]
- I_{CaL} ▲ [3]
- I_{NaCa ≜} [4]

- 1. Bassani R.A., (2006) Braz J Med Biol Res 39(3): 393-403
- 2. Taffet, G.E., et al., (1993) Am J Physiol Heart Circ Physiol; 264(5): p. H1609-H1614.
- 3. Isenberg, G., et al., (2003) Cell Calcium. 34(3): p. 271-280.
- 4. Wong, K., et al., (1998) Cardiovasc. Res. 37, 115–122.

The sensitivity for parameter modulation





0% 20% 50% 70%

$$\mathbf{I_{CaL}} \uparrow \quad \mathbf{I_{NaCa}} \uparrow \quad \mathbf{I_{to}} \downarrow \quad \mathbf{V_{max}} SERCA \downarrow$$

Building an Aged Population

- 60 age-related sets
- normal distributions for parameter deviation
 - SD 0.1
 - Mean:
 - 1.25 for g_CaL and K_NaCa
 - 0.85 for g_to and
 V_max SERCA



Arrhythmogenicity in aging

For each aged population:

- Detect the EAD cases
- Identify models with a prolonged AP >10%
- Calculate the score

$$W = \frac{N_{EAD}}{N_{total}}$$



Linear model

 $score = w_0 + w_1 g_{CaL} + w_2 P_{NaCa} + w_3 g_{to} + w_4 V_{max_{up}}$

coefficient	value
w _o	-0.92
W ₁	0.71
W ₂	0.07
W ₃	0.03
W ₄	-0.25

Determination coefficient = 0.986

Model parameters distribution

blue: subpopulation without changes in AP

orange: subpopulation vulnerable to EAD after aging

green: subpopulation vulnerable to AP prolongation after aging





Conclusion

- Age-related changes in the cellular ionic currents may lead to an arrhythmogenic increase in APD and the emergence of EAD in human cardiomyocytes.
- The population based analysis allows one to classify age-related effects on different ECC mechanisms.
- An increase in the I_CaL current and a decrease in the SERCA flow is shown to be the significant factors of age-related cellular remodeling.
- An upregulation of I_CaL, IKs, IKr and downregulation of SERCA uptake increase risks of arrhythmias

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