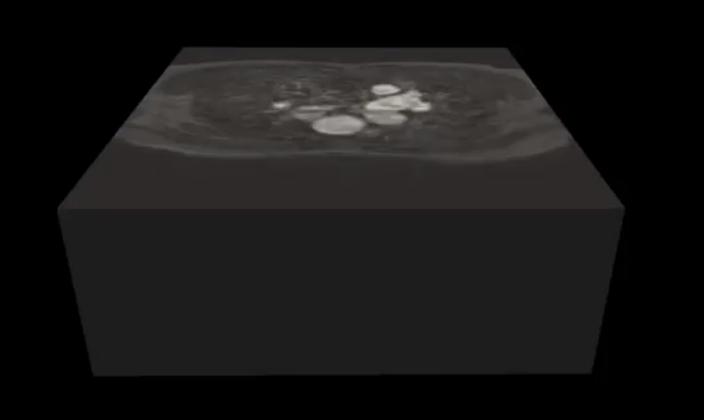
CONSTRUCTING AN ACTION POTENTIAL MODEL: EASY AS ABC?

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4 Sep 2017



IN-SILICO MODELLING OF CARDIAC ELECTROPHYSIOLOGY



Computational modelling important test hypotheses. Homogenous tissue. Ignore cell-level propagation	1.

TOWARDS A DISCRETE-CELL APPROACH

Microscopic propagation can have an effect. Discrete-cell modelling. Separate interconnected regions. Validation difficult. HL1-6. First step.	n

MODELLING AN ACTION POTENTIAL

$$I_{K} = G_{K} \cdot a \cdot i \cdot (V - E_{K})$$

$$\frac{da}{dt} = \frac{a_{ss} - a}{\tau_{a}} \qquad \tau_{a} = 0.493e^{-0.0629V} + 2.058$$

- Parameters fit to experimental patch clamp data.
- Traditional fitting methods do not account for uncertainty in estimates.
- Approximate Bayesian Computation (ABC) produces posterior distribution for each parameter.



HYPOTHESIS

The ABC approach can be used to construct a validated mathematical model of the action potential of a HL1-6 cell while taking into account uncertainties in parameter estimates arising from insufficient fitting data, biological variability and/or parameter redundancy.

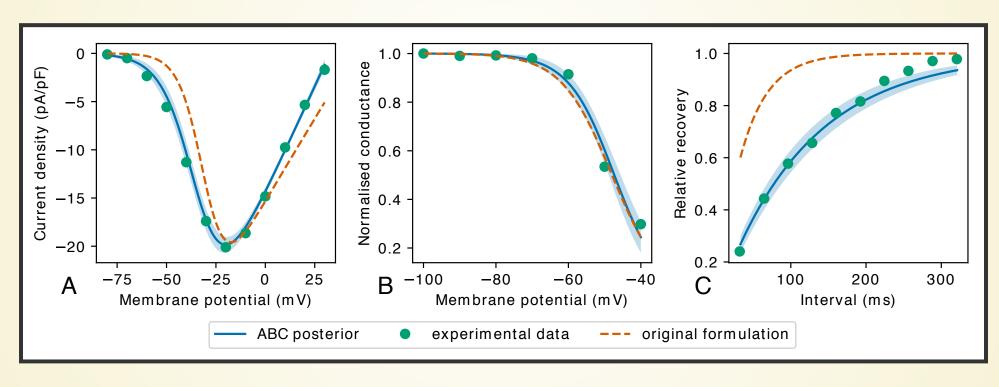
AIMS

- Develop an ABC implementation to estimate parameter posterior distributions for individual ion currents.
- Investigate the sources of any uncertainty and unidentifiability in parameter estimates.
- Construct the full action potential model and validate with action potential recordings from biological experiments.



AIM 1: PARAMETERISING ION CURRENTS USING ABC

 ABC outperformed traditional maximum likelihood estimation, and provides uncertainty in simulation output.

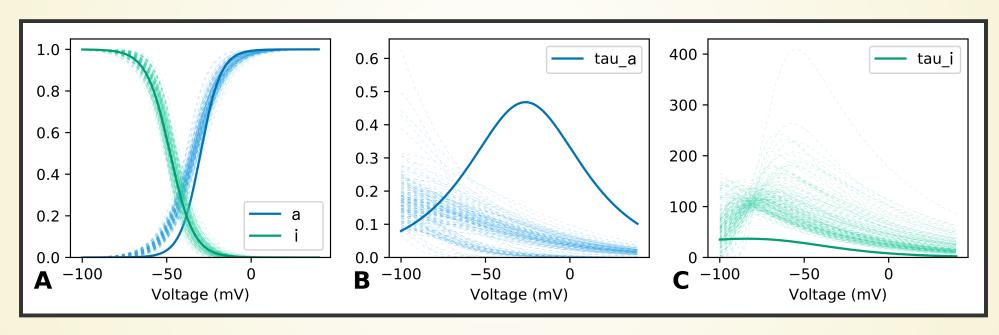


ABC vs Maximum Likelihood fitting of T-type Calcium current: A = activation, B = inactivation, C = recovery.

ee graphs/protocols. Graph description. Improves fit. Distribution narrow. Recovery harder to fit. Appears well strained but underlying variables.	

AIM 2: INVESTIGATING UNIDENTIFIABILITIES

 Time constant curves could not be constrained by standard protocol patch clamp data.

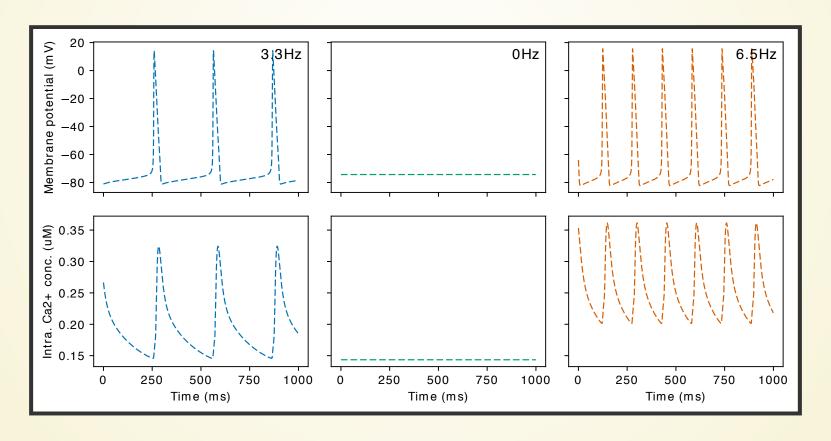


Underlying variables in ion current equations with ABC parameter posteriors: A = steady-state, B = activation time constant.

Plot description. Steady-state well constrained. Time constants wide variation. Practical unidentifiability. Rerun ABC before full model.

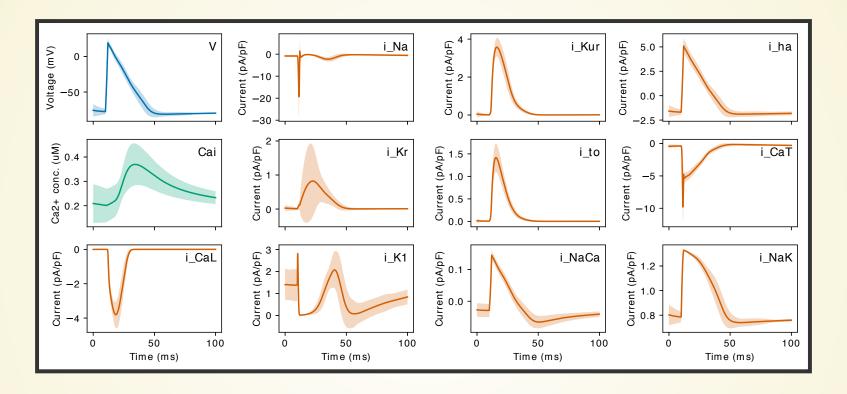
AIM 3: FULL ACTION POTENTIAL MODEL

- Automaticity in 56% of runs with mean firing rate 4.9±2.0 Hz.
- Comparable qualitative action potential and Calcium transients between simulations and experiments.



Random draw parameters different output. 200 sims 1 second. Automaticity. Qualitatively similar.

AIM 3: INITIAL VALIDATION



variable	HL1-6 model	HL1-6 experiment	HL-1 experiment
APD ₉₀ (ms)	$\textbf{32.4} \pm \textbf{0.5}$	42 ± 9	
V_{rp} (mV)	-77.5 ± 0.3	-67 ± 2	-68.8 ± 1.6
AP amplitude (mV)	96.8 ± 0.5	105 ± 2	
V _{overshoot} (mV)	19.3 ± 0.5		15.3 ± 1.9

Advantage propagate uncertainty. Quantitative meathreshold.	asurement. Re	esting potential o	difference. So	dium channel fitting	

CONCLUSIONS

- ABC is an effective approach to infer model parameters while accounting for uncertainties and/or unidentifiabilities.
- Standard protocol voltage patch clamp data is not sufficient to completely constrain time constant parameters.
- The full action potential model reproduces qualitative and quantitative characteristics of the HL1-6 myocyte.

Hypothesis partially accepted. Aim 1 complete. Time constants not constrained. Part validation. Further work.

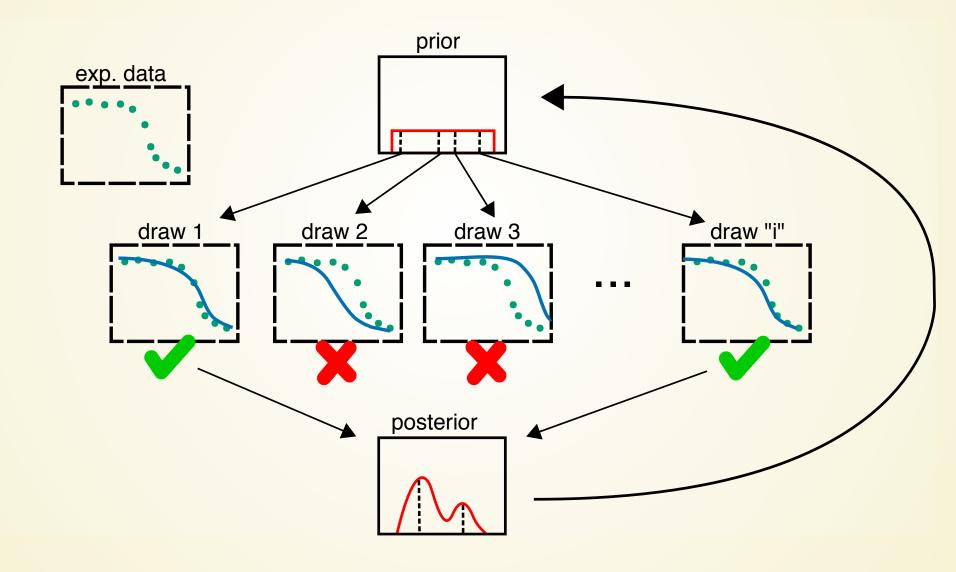
ACKNOWLEDGEMENTS

- Dr Chris Cantwell
- All members of ElectroCardioMaths Programme
- British Heart Foundation





APPROXIMATE BAYESIAN COMPUTATION (ABC)



MODEL OVERVIEW

Simulate by solving 35 ODEs at each time step.

