

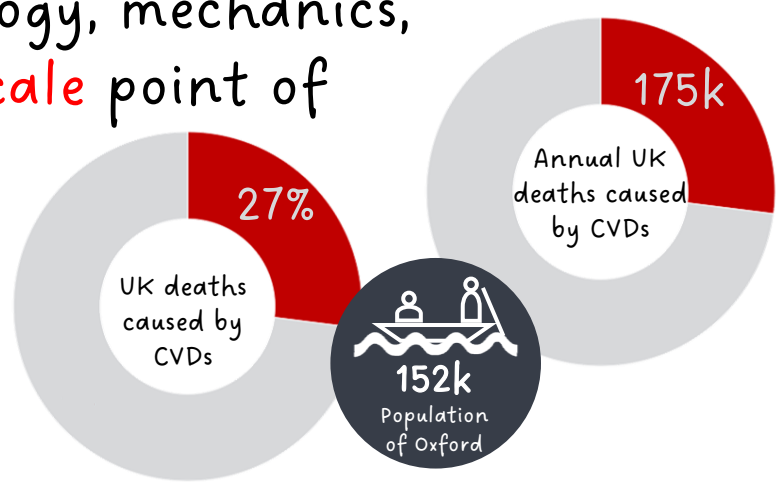
Mathematical models to create cardiac digital twins for precision healthcare



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THE PROBLEM

- There are around 7.6 million people living with cardiovascular diseases (CVDs) in the UK
- CVDs cause 27% of all deaths in the UK - one every 3 minutes
- The specific causes require studying the heart from a **multiphysics** (electrophysiology, mechanics, fluid dynamics) and **multiscale** point of view (from molecules to whole-organ dynamics)

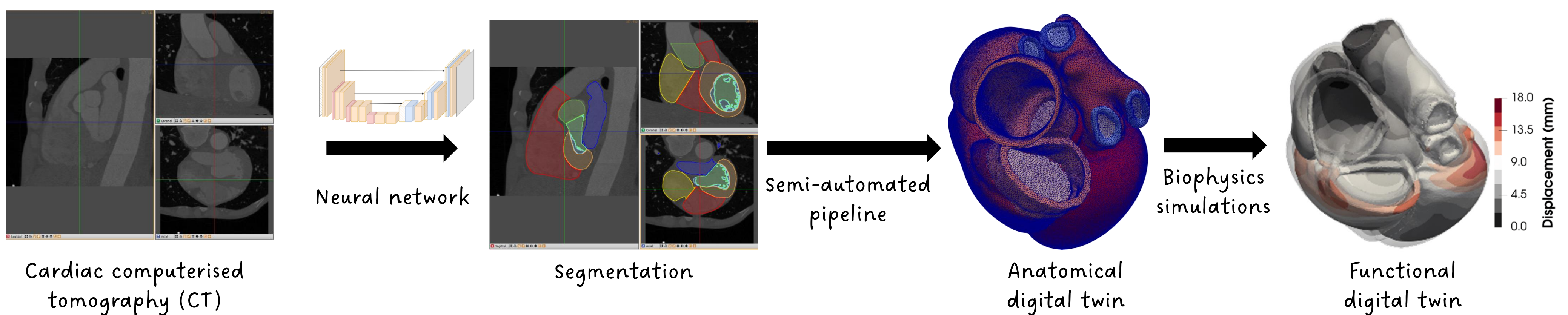


HOW CAN CARDIAC DIGITAL TWINS HELP?

- A **cardiac digital twin** (DT) is a virtual representation of the heart of a person
- This representation includes patient-specific data and the known biology and physics of the heart
- DTs allow us to run **simulations** of a patient's condition and test multiple treatments
- Create a DT of patient can take months because of the complexity of the process and the number of **parameters** involved in the biophysics equations

Electrophysiology	Mechanics	Blood flow
$\ \nabla \tau(x)\ _V = \sqrt{\nabla \tau(x)^T \nabla \tau(x)} = 1 \quad \forall x \in \Omega$	$\Psi(C) = \frac{\kappa}{2} (\ln J)^2 + \frac{\alpha}{2} (e^Q - 1)$	$\frac{d^2 V(t)}{dt^2} = A_1 \cdot V(t) + A_2 \frac{dP(t)}{dt} + A_3 P(t)$
$V(x) \cong CV_f^2 (f(x) \otimes f(x)) + CV_s^2 (s(x) \otimes s(x))$	$Q = b_f E_{ff}^2 + b_t (E_{ss}^2 + E_{nn}^2 + 2E_{sn}^2) + 2b_{fs} (E_{fs}^2 + E_{fn}^2)$	$A_1 = \frac{1}{Z \cdot C} + \frac{1}{R \cdot C}, \quad A_2 = -\frac{1}{Z}, \quad A_3 = -\frac{1}{Z \cdot R \cdot C}$

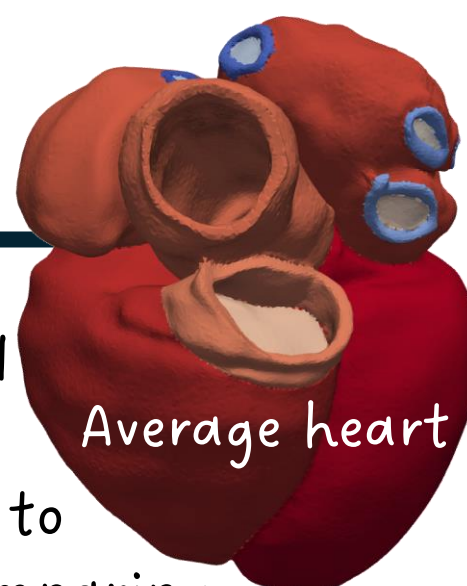
FROM MEDICAL IMAGES TO DIGITAL TWINS



We applied this pipeline to 20 healthy subjects and made **openly available** the first cohort of healthy whole-heart models

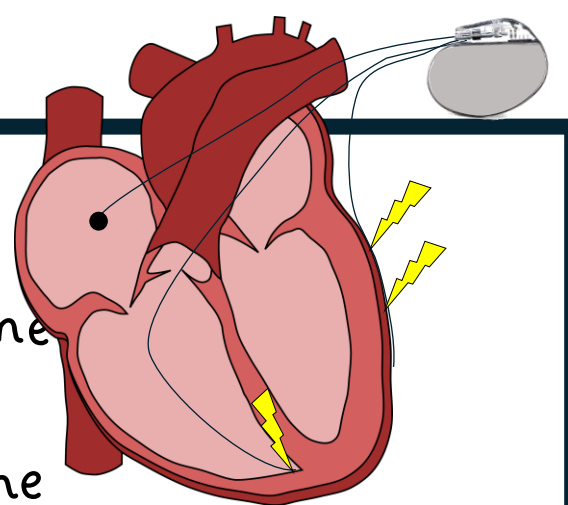
TAKING SHAPE INTO ACCOUNT

- There is a poor understanding of how specific anatomical changes affect different cardiac outputs
- Statistical shape modelling** is a mathematical technique to describe complex shapes with a limited set of numbers comparing them to a generated "average" shape
- We applied this technique to our cohort of healthy hearts and used those numbers to **link** localised changes in **anatomy** to changes in the simulation of the cardiac **function**
- We found how small changes in the shape can indicate early indications of hypertension
- We made open-source a synthetic cohort of more than 1000 heart models generated using this technique



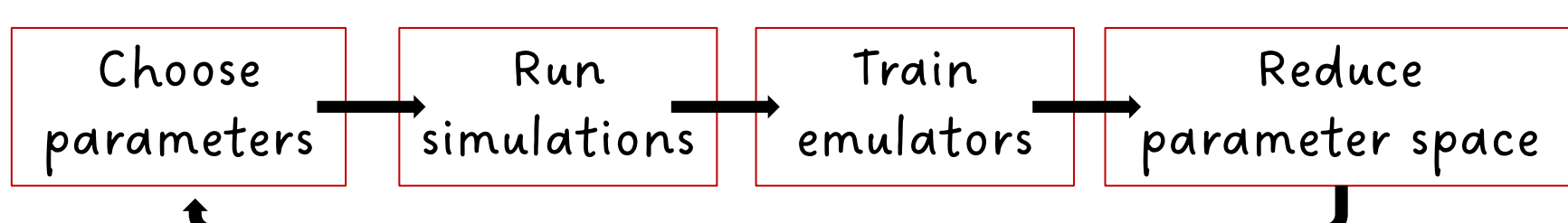
IN-SILICO TRIAL OF MEDICAL DEVICES

- Dissynchronous heart failure (HF) is a condition where the two sides of the hearts do not beat at the same time
- Cardiac resynchronization therapy (CRT) is one of the main treatments, where a pacemaker is inserted and the heart is shocked at several locations
- Where to install the CRT electrodes can change between patients and it is not clear if there is an optimal location
- We created DTs of 24 HF patients and 20 healthy ones representing recovered patients
- We tested more than 8000 pacemaker configurations, representing a **virtual (or in-silico) trial**
- We found that **differences in anatomy was more relevant than differences in the electrodes' locations**



ACCELERATING CLINICAL TRANSLATION

- One of the main bottlenecks with DTs is **parameter fitting**
- Parameter fitting consists of finding the set of parameters that describes best given clinical data
- We developed a pipeline based on **Bayesian History Matching** and **Gaussian Process Emulators** to find a reduced parameter space
- We can then **reuse** that **information** with new patients reducing drastically the computational load to create DTs



NOW WHAT?

- New modalities, like magnetic resonance imaging (MRI)
- New diseases, like hypertrophic cardiomyopathy (HCM)
- Scale up: can we do hundreds of DTs in a sensible timeframe?
- How can we update DTs with new visits to the hospital?
- Can we add the explicit effect of hormones?

THE PAPERS

