

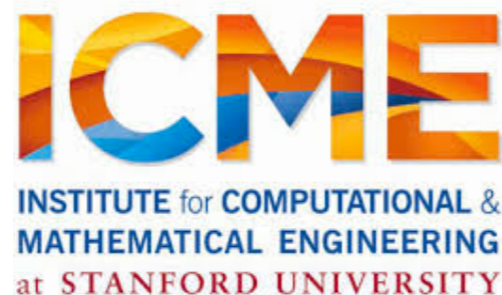
Computational Methods for Personalized Medicine in Cardiovascular Disease

Alison L. Marsden, PhD
Associate Professor

Departments of Bioengineering and Pediatrics, and by
courtesy, of Mechanical Engineering
Institute for Computational and Mathematical Engineering
Stanford University



Stanford
MEDICINE

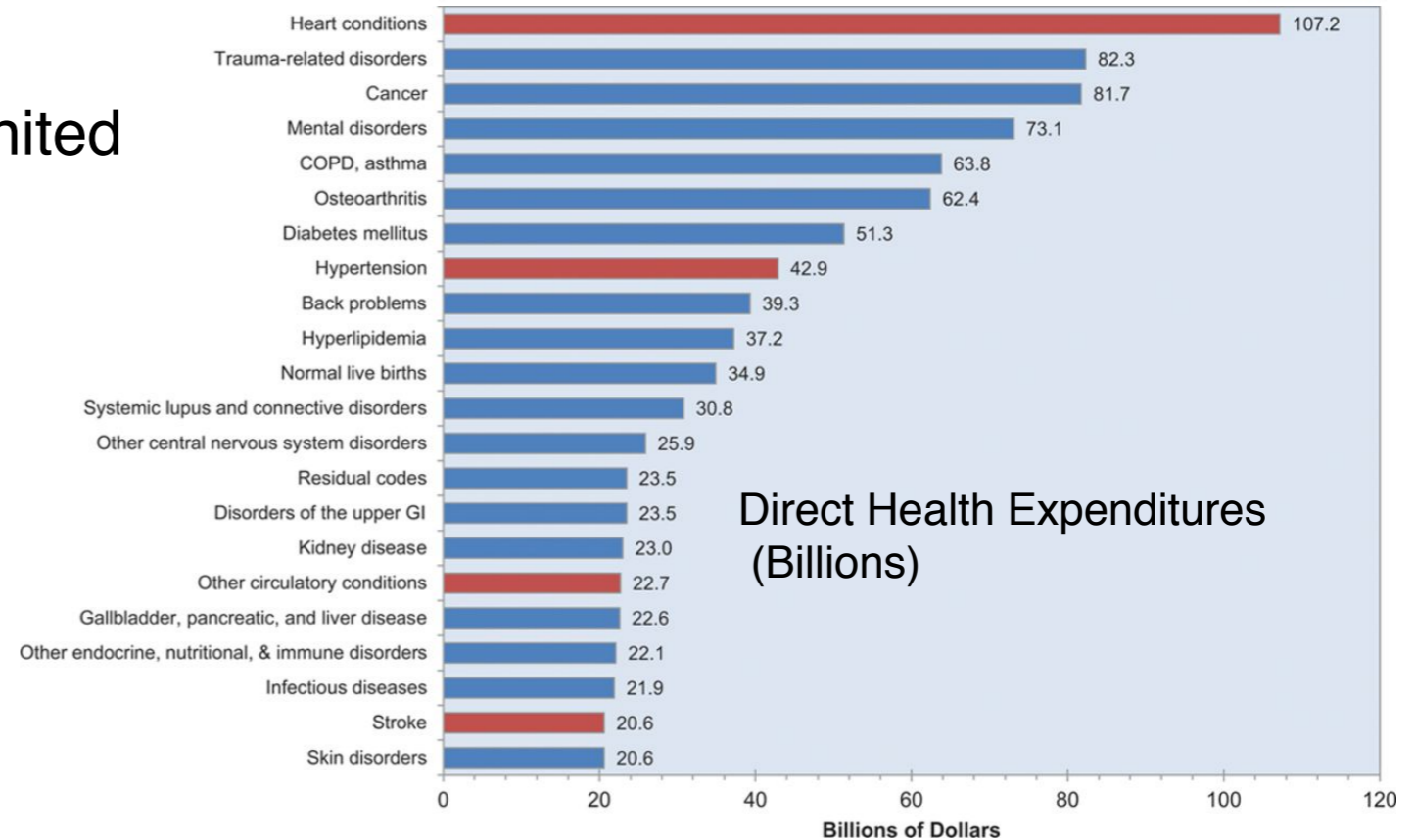
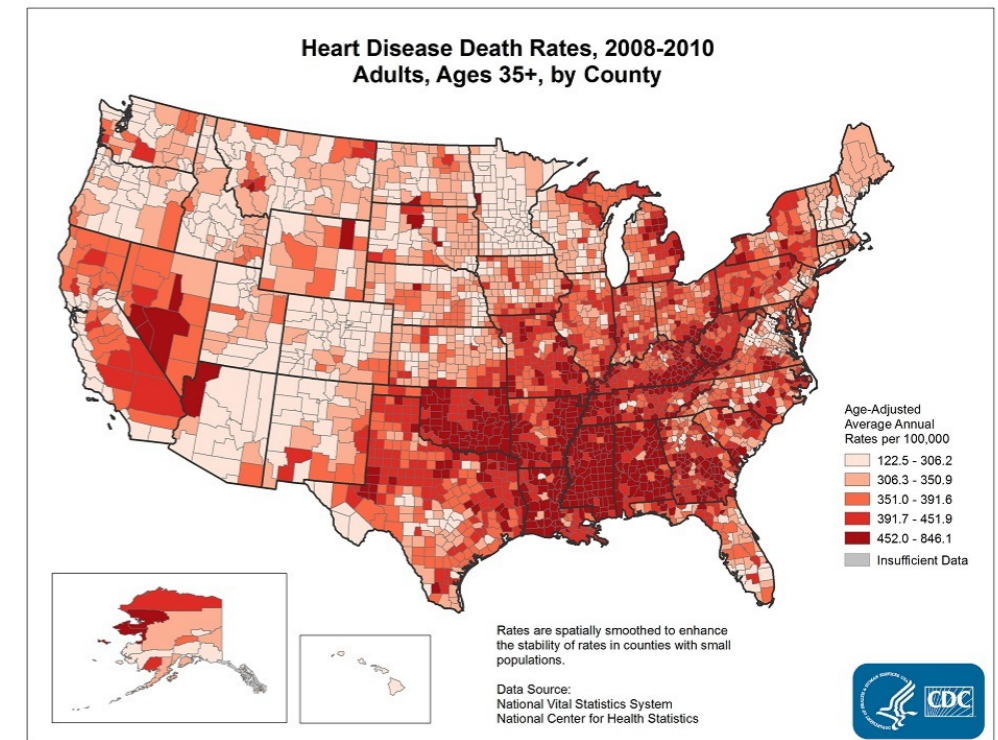


Lucile Packard
Children's Hospital
AT STANFORD

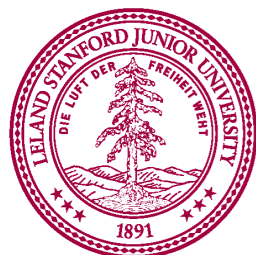


Cardiovascular Disease

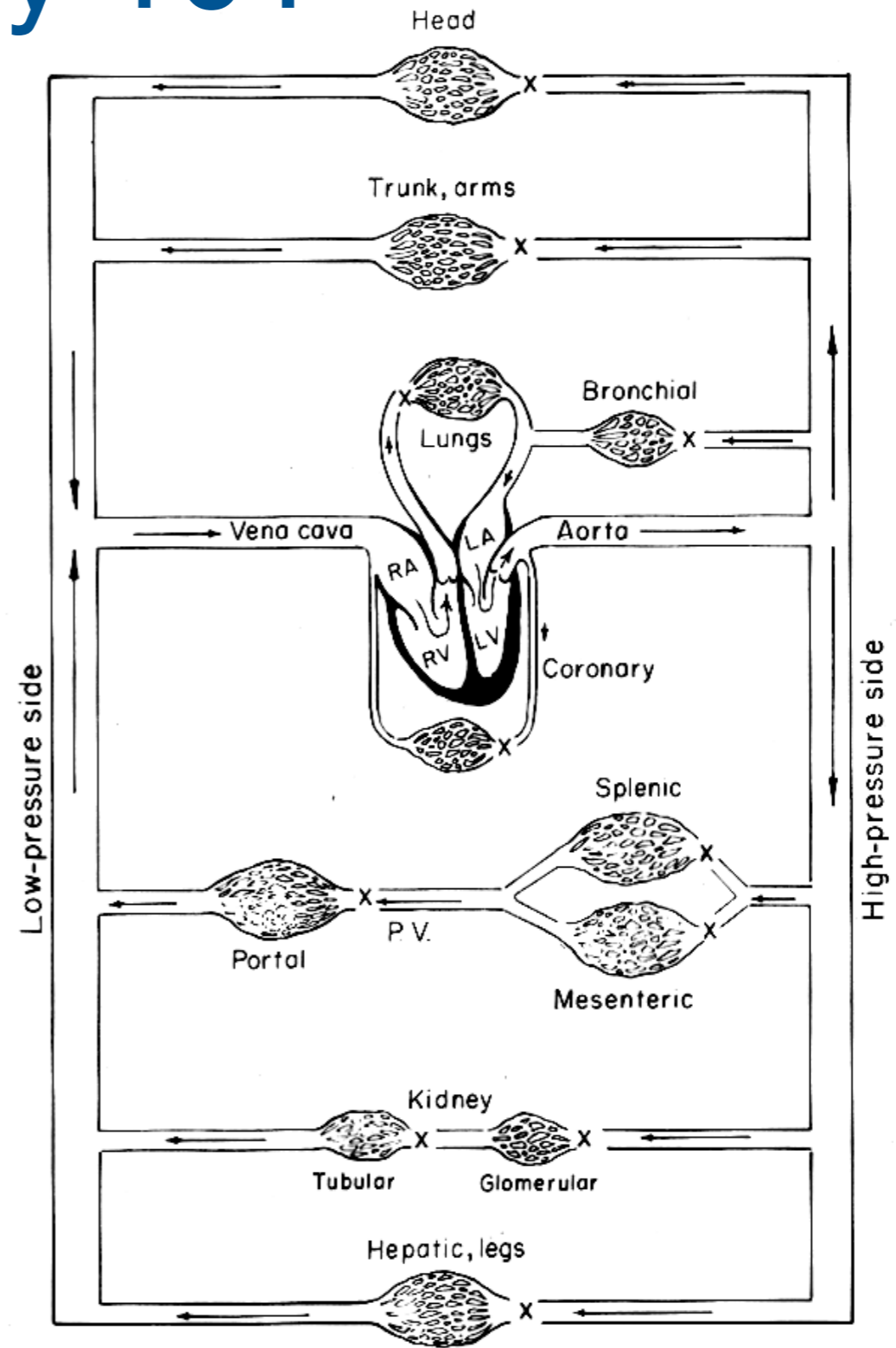
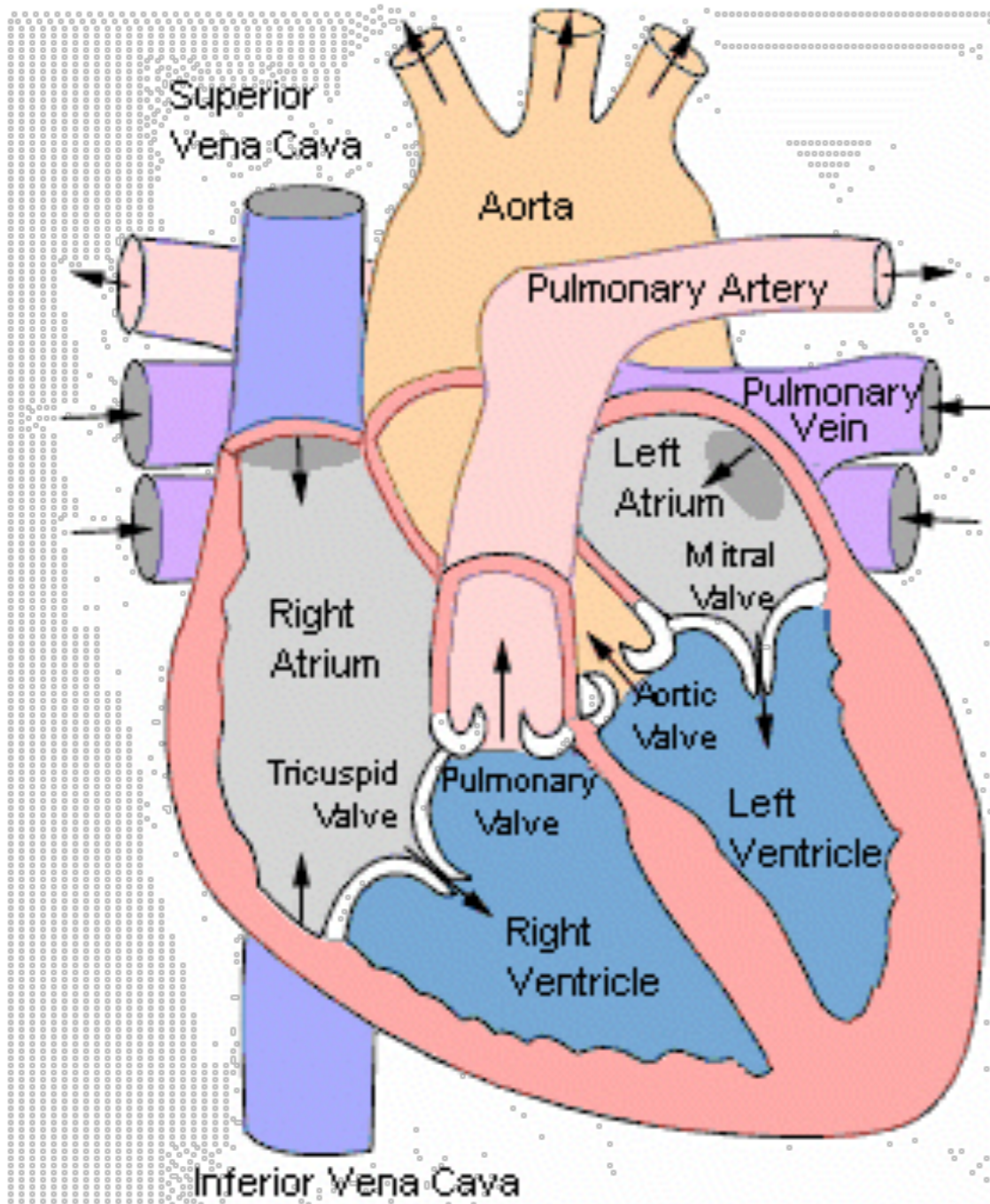
- CVD is the **leading cause of death for men and women globally** ~17 Million deaths worldwide
- ~ **600,000 Americans** die from heart disease each year ~ **1 in every 4 deaths**
- Cardiovascular disease costs the United States **>\$500 billion** each year
- 1/100 children are born with a **congenital heart defect**



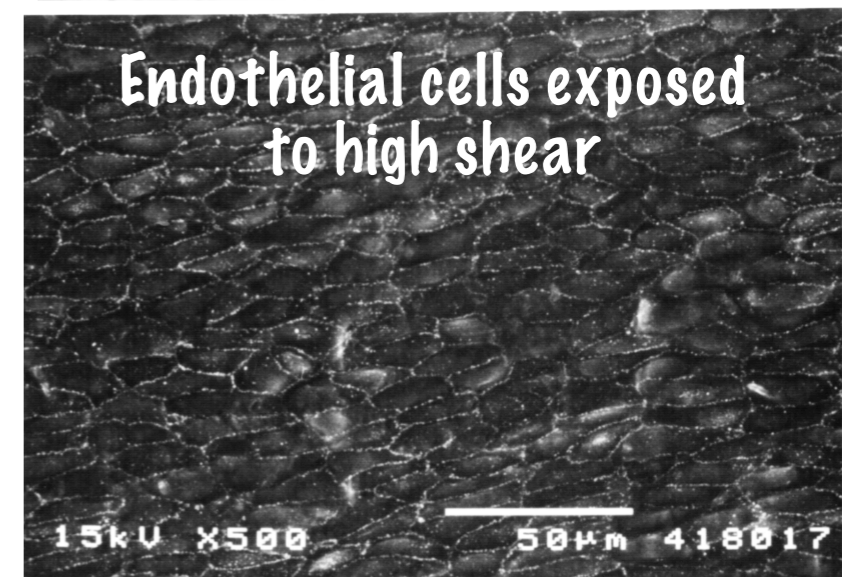
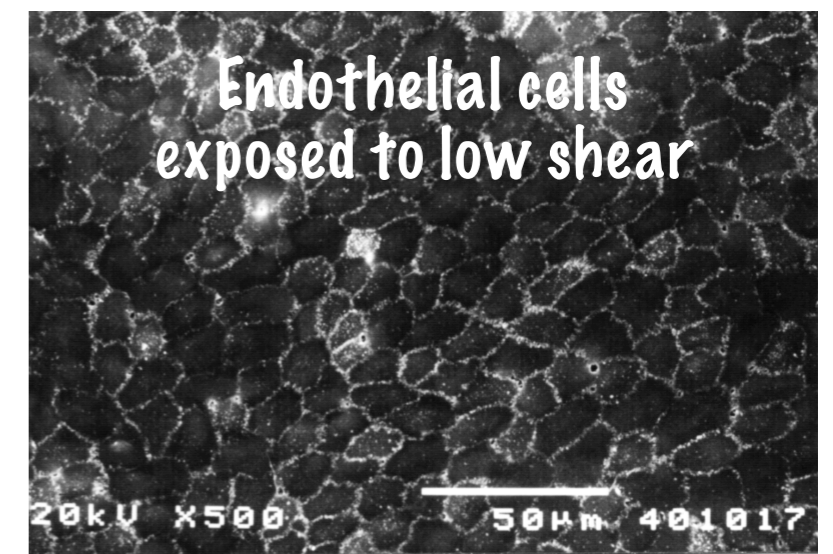
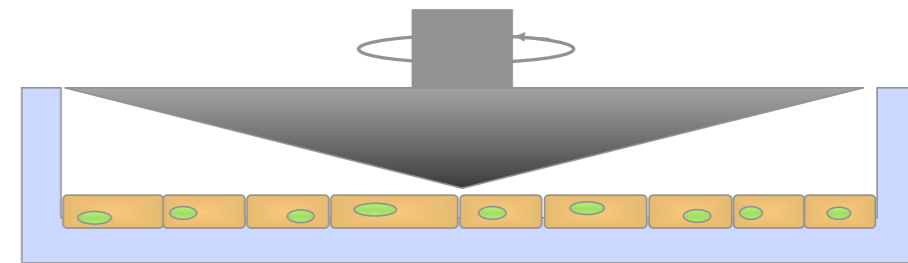
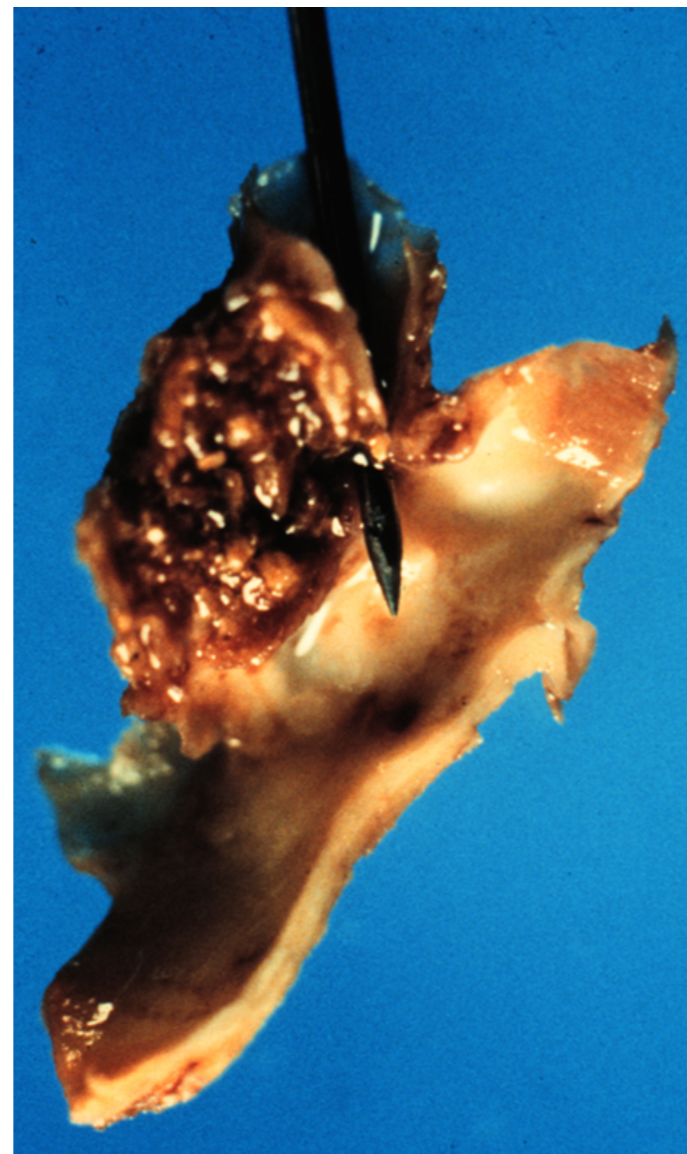
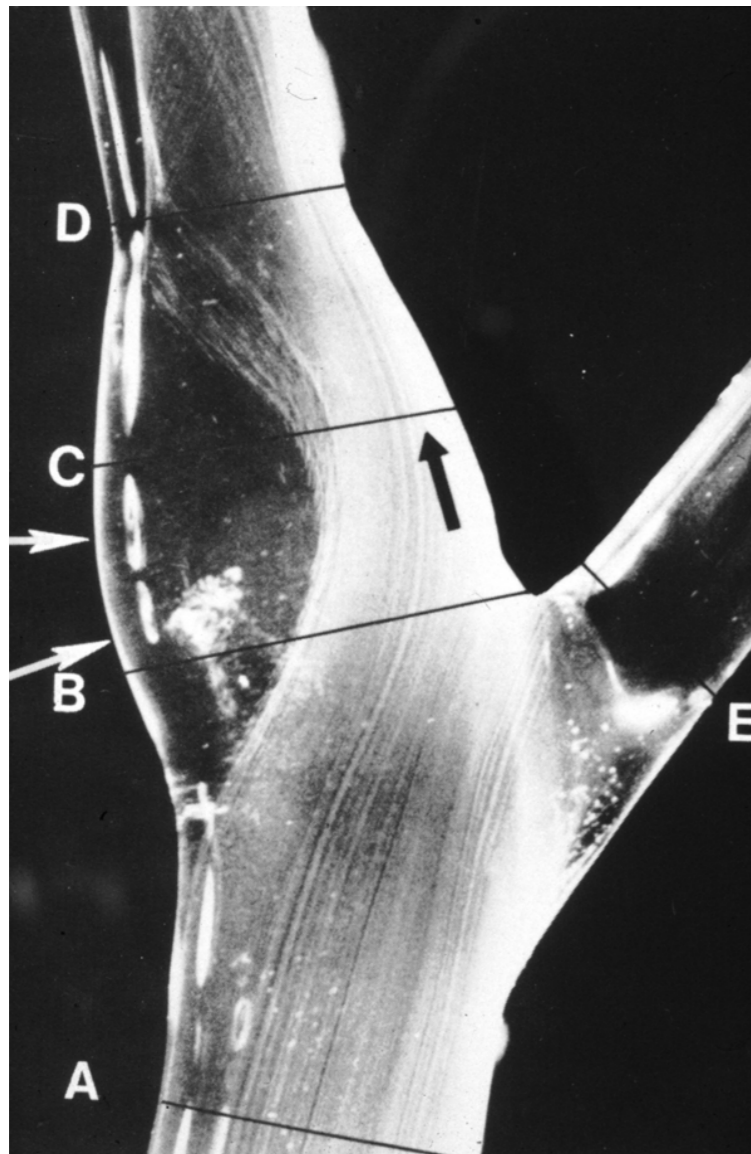
Centers for Disease Control and Prevention
AHA 2014 Statistics Report, Circulation



Anatomy 101



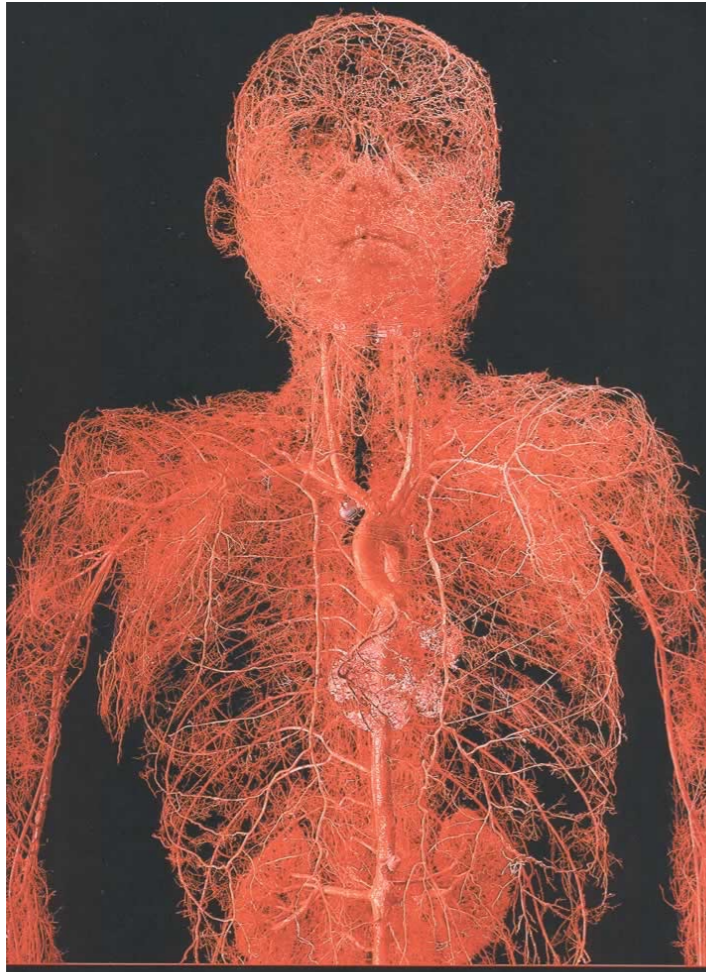
Mechanical Forces Alter Disease Progression



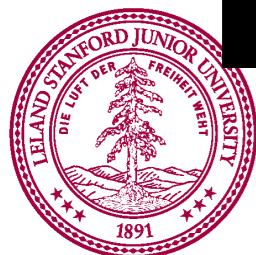
Glagov S., Zarins C.K., Giddens D.P., Ku D.N. (1988) Establishing the Hemodynamic Determinants of Human Plaque Configuration, Composition and Complication. Role of Blood Flow in Atherogenesis. Springer.



Cardiovascular System Complexity



- Cardiovascular system is comprised of billions of blood vessels
 - Most are beyond the limits of our imaging resolution
- Transport and delivery system for oxygen, nutrients, hormones
- Adaptive and dynamic
 - Fluid and solid mechanics
 - Biological response - mechanobiology
 - Physiology
- Health and disease models



Motivation for Cardiovascular Modeling

Virtual surgery and treatment planning

Fundamental disease mechanisms

Patient risk stratification

Personalized Medicine

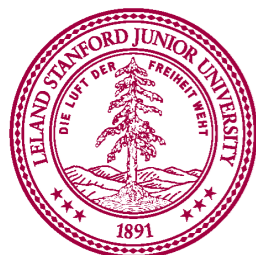
Test novel surgical concepts

Augmenting clinical imaging

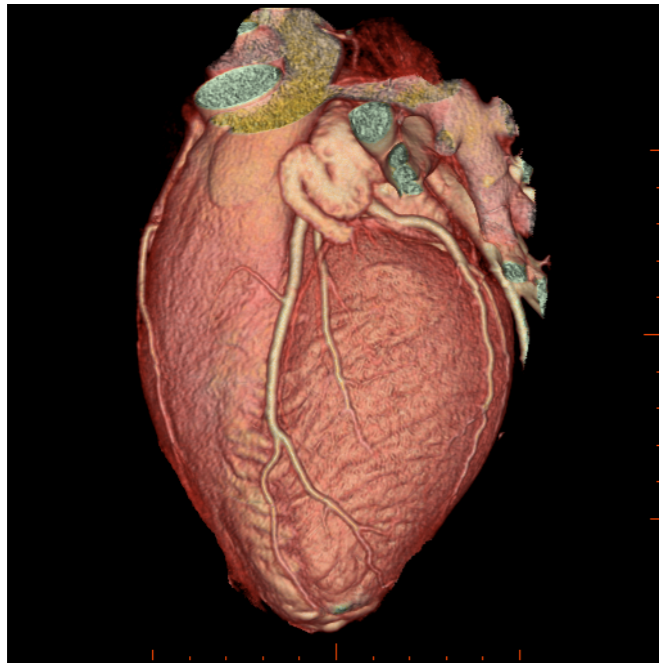
Vascular Growth and Remodeling

Prediction of post-surgical conditions

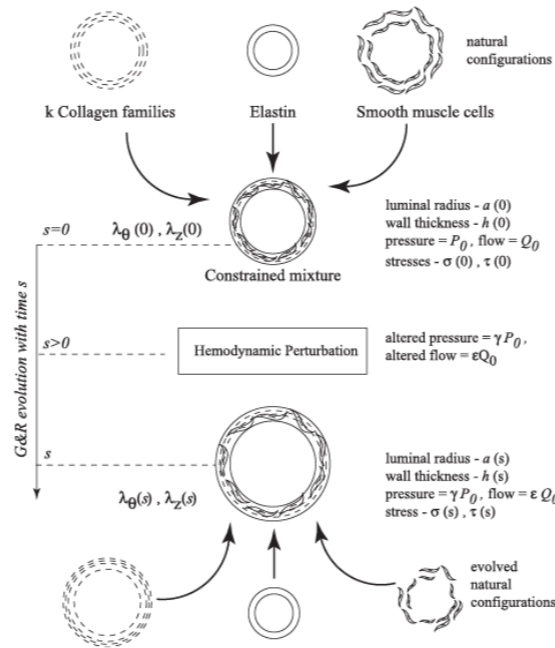
Mechanobiology



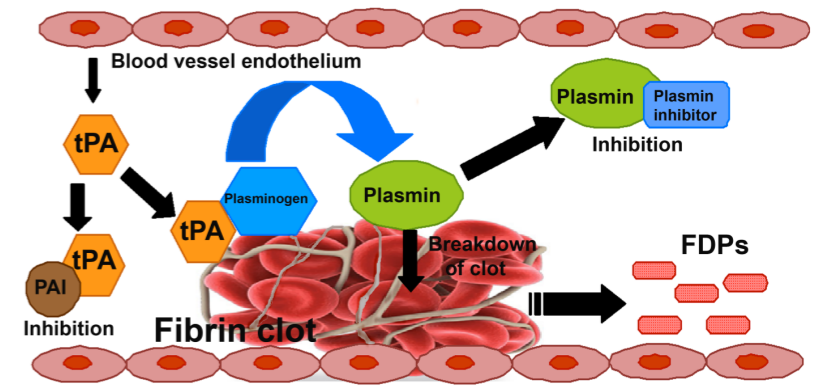
Cardiovascular Modeling: Physics



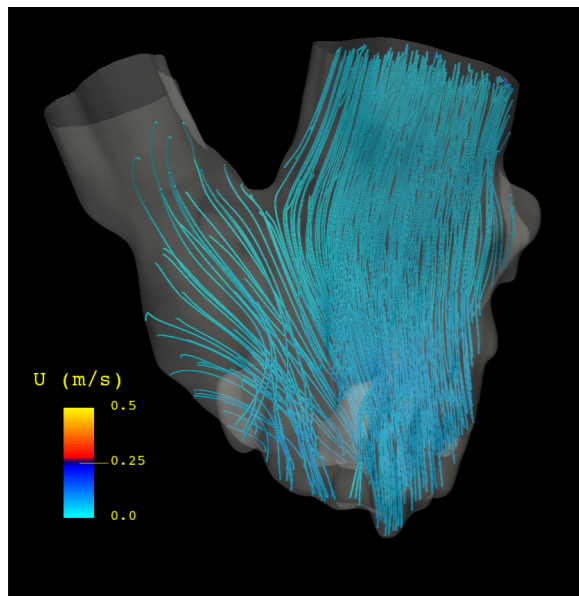
Multi-physics, Multi-scale



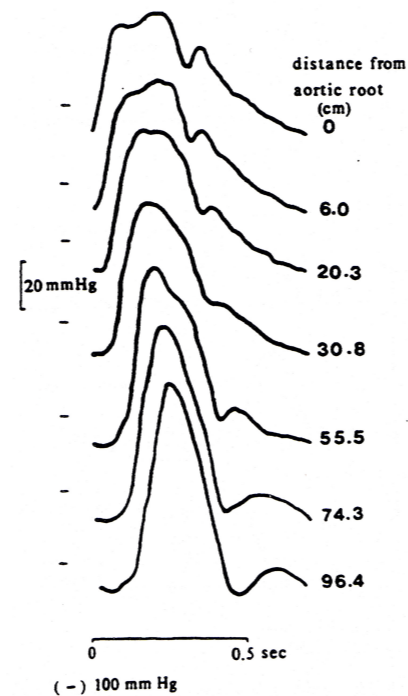
Vascular Mechanobiology



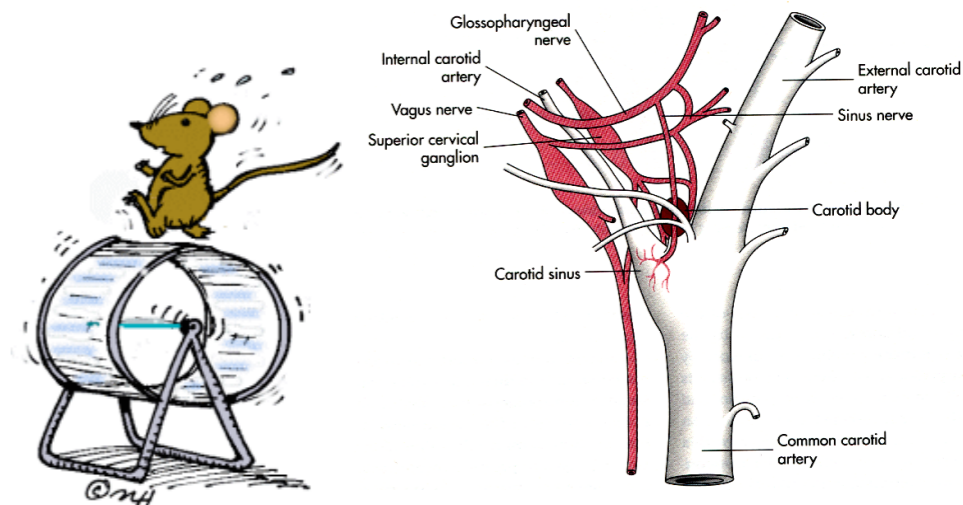
Thrombolysis and Biochemistry



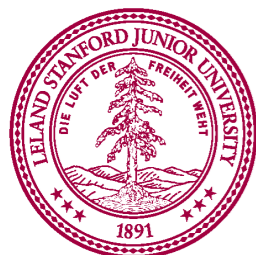
Fluid-structure interaction



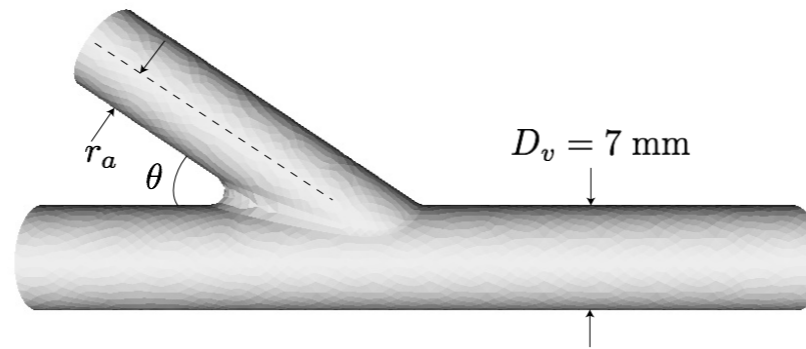
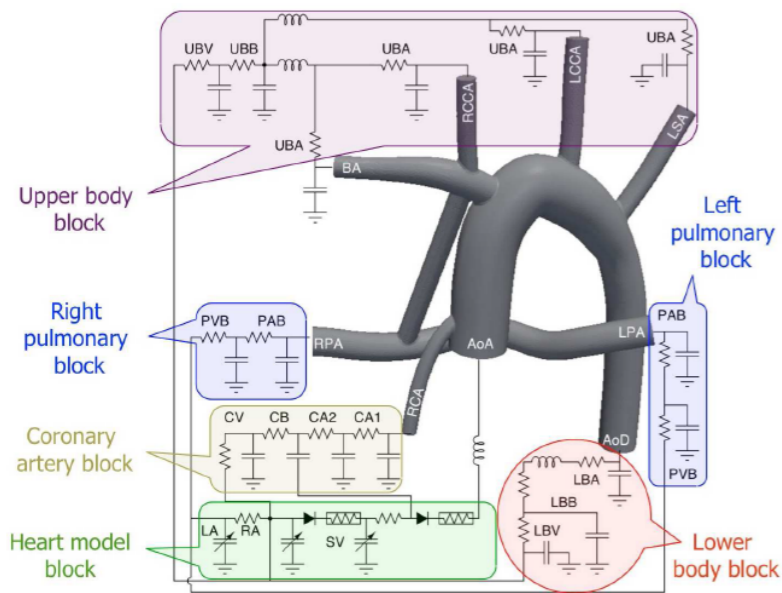
Wave Propagation



Autoregulation and Adaptation



Cardiovascular Modeling: Tools

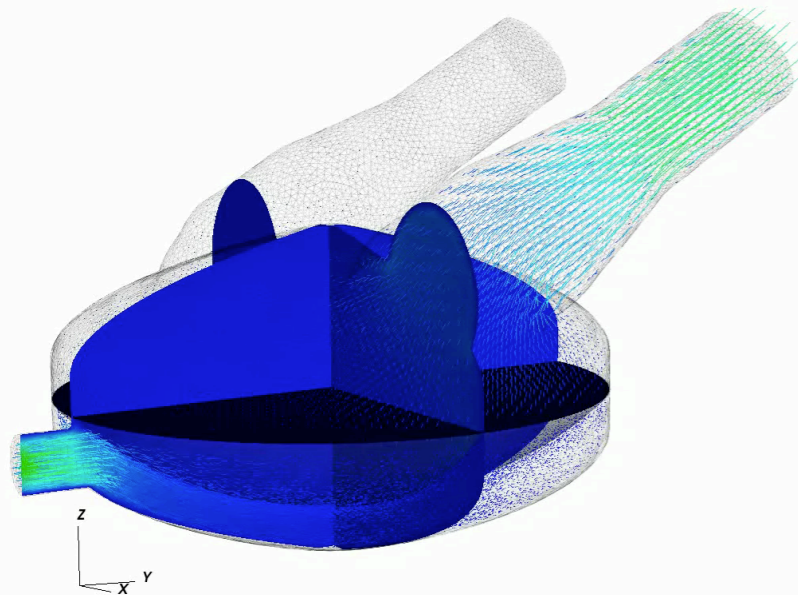


Shape Optimization

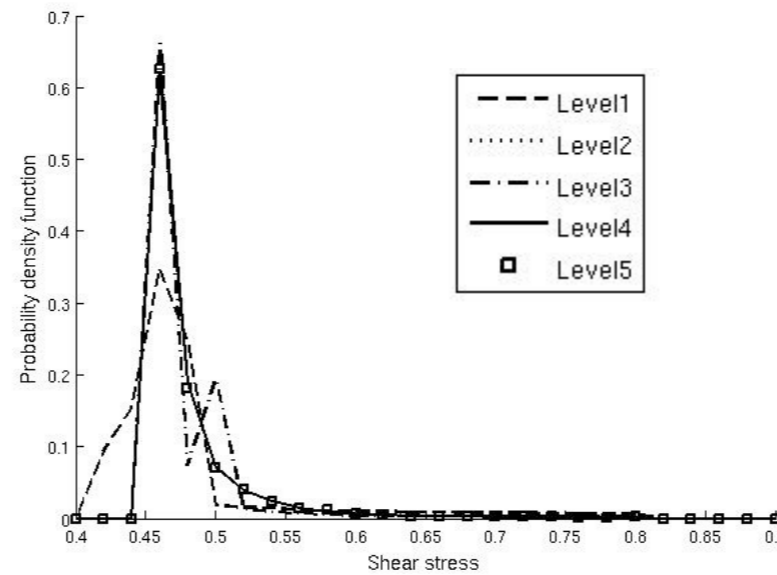


Predicting Clinical Outcomes

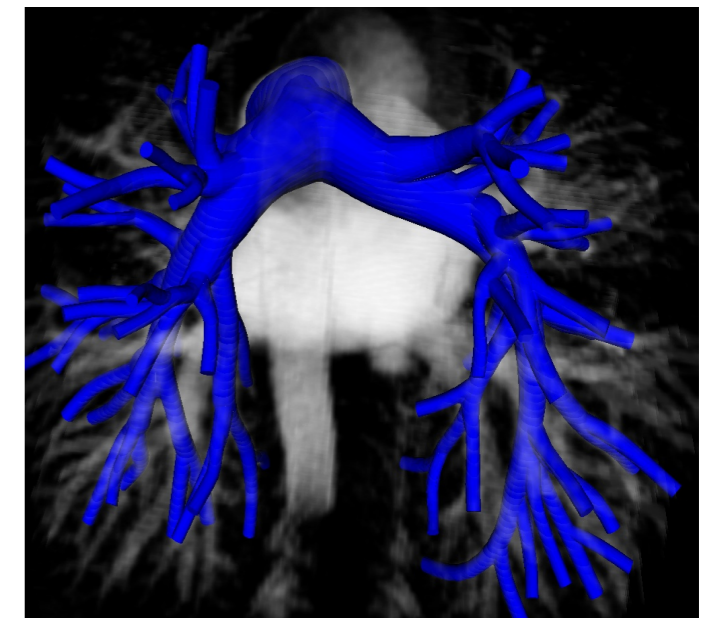
Physiology and clinical data



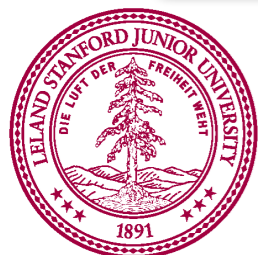
Medical Device Design



Uncertainty Quantification



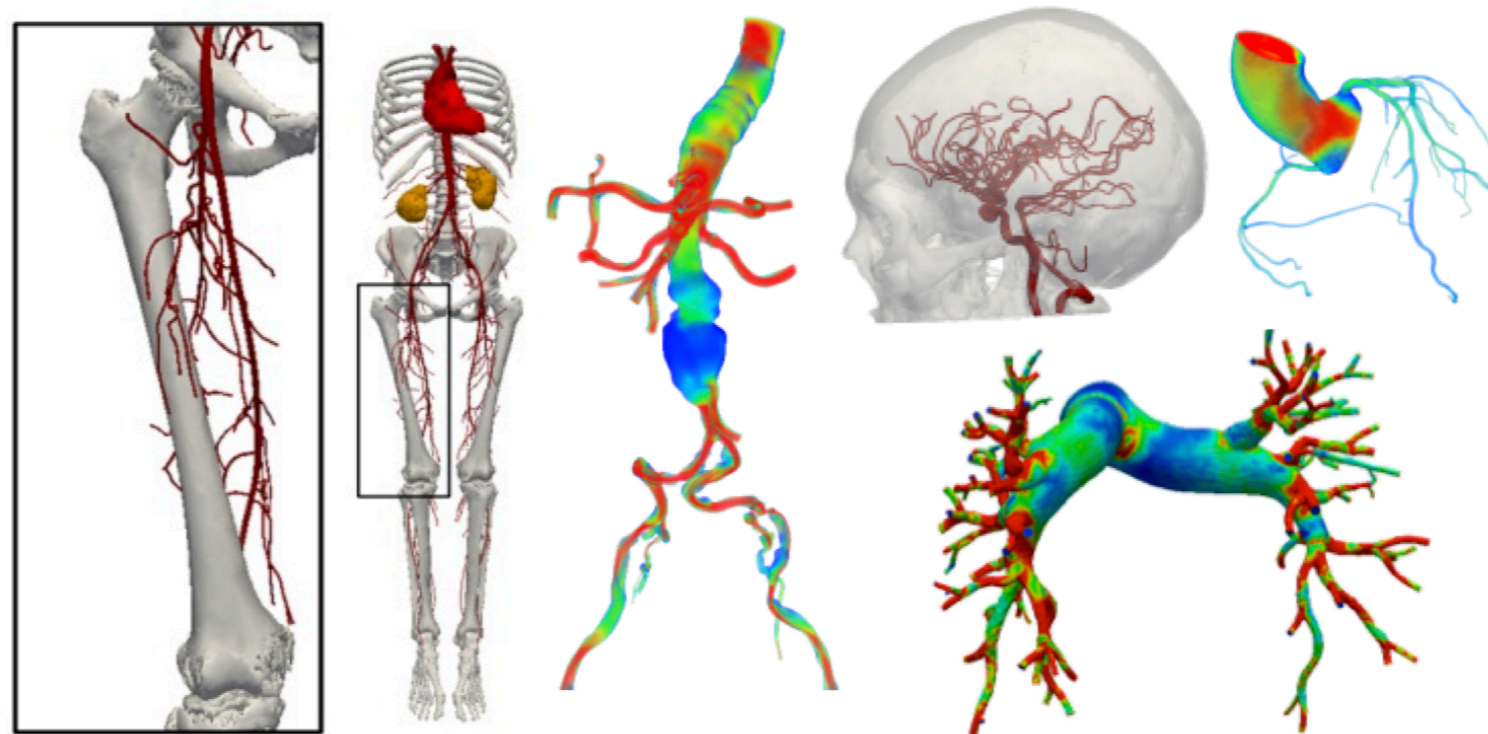
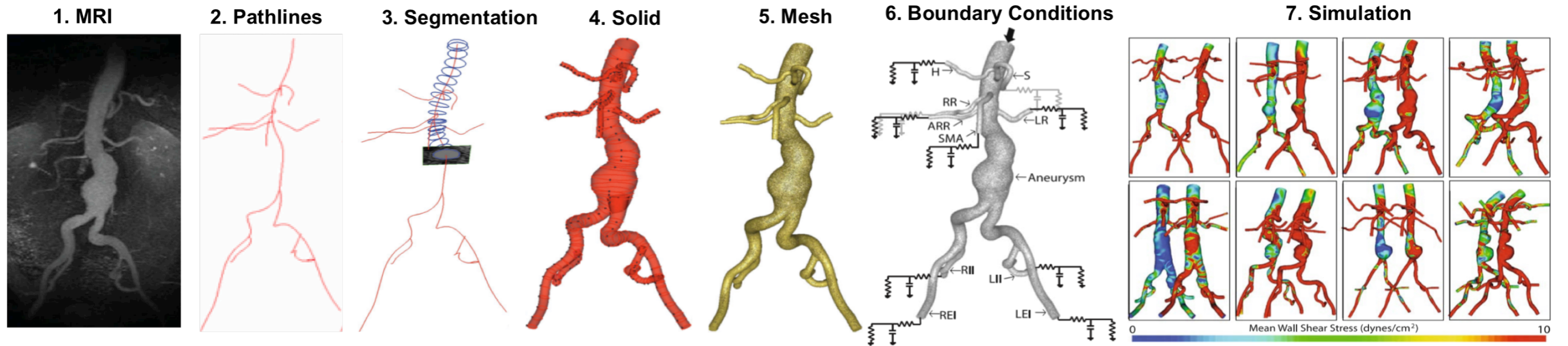
Machine Learning for Image Segmentation



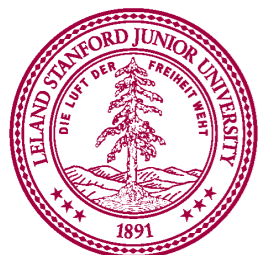
Computational Methods

Making Patient-Specific Predictions

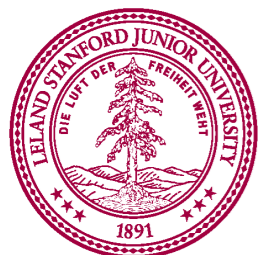
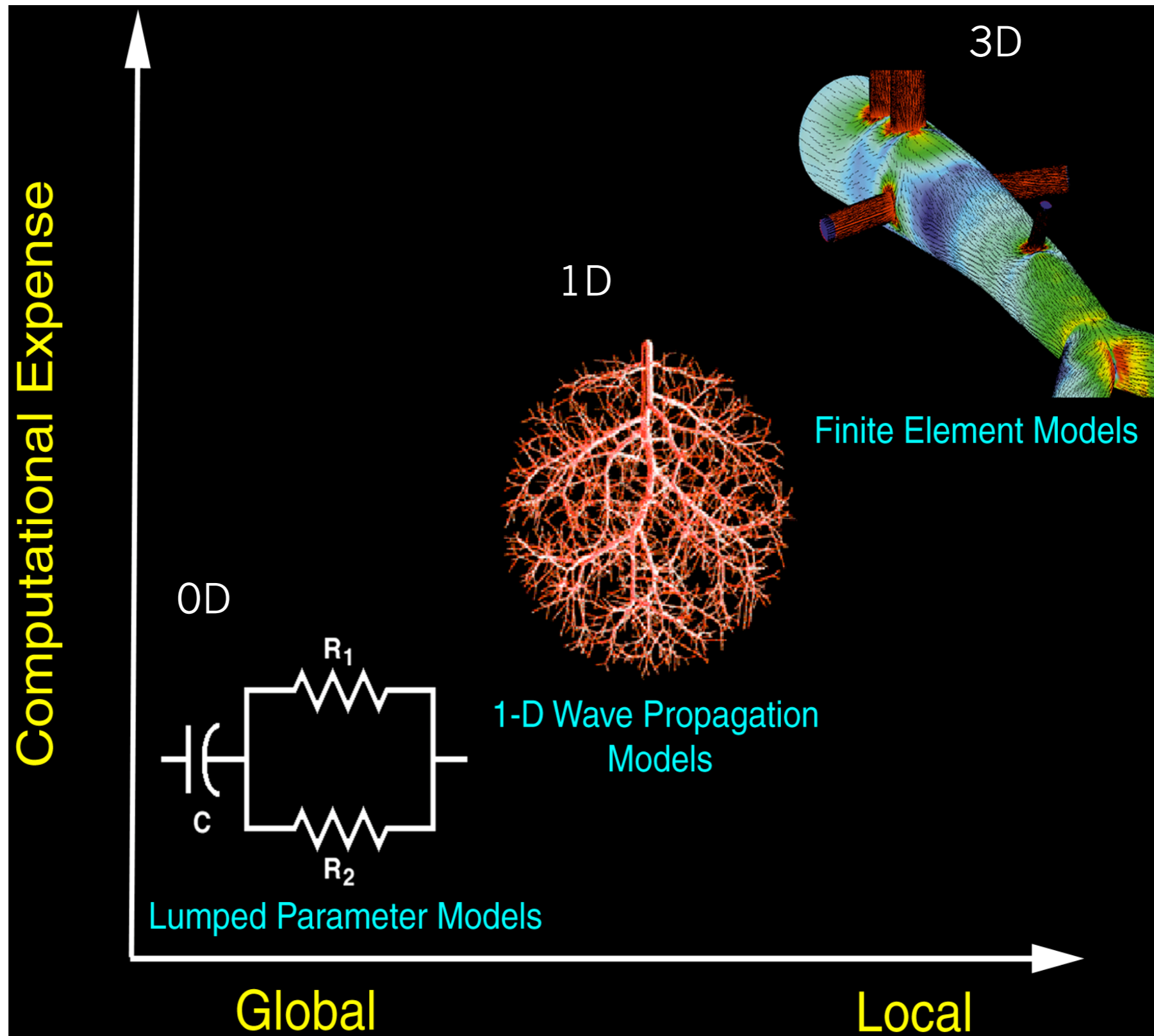
Patient-Specific Modeling



Updegrave, A., Wilson, N.M., Merkow, J., Lan, H., Marsden, A.L., Shadden, S. C., "SimVascular - An Open Source Pipeline for Cardiovascular Simulation," *Annals of Biomedical Engineering*, Vol 45 (3), pp. 525-541, (2017).



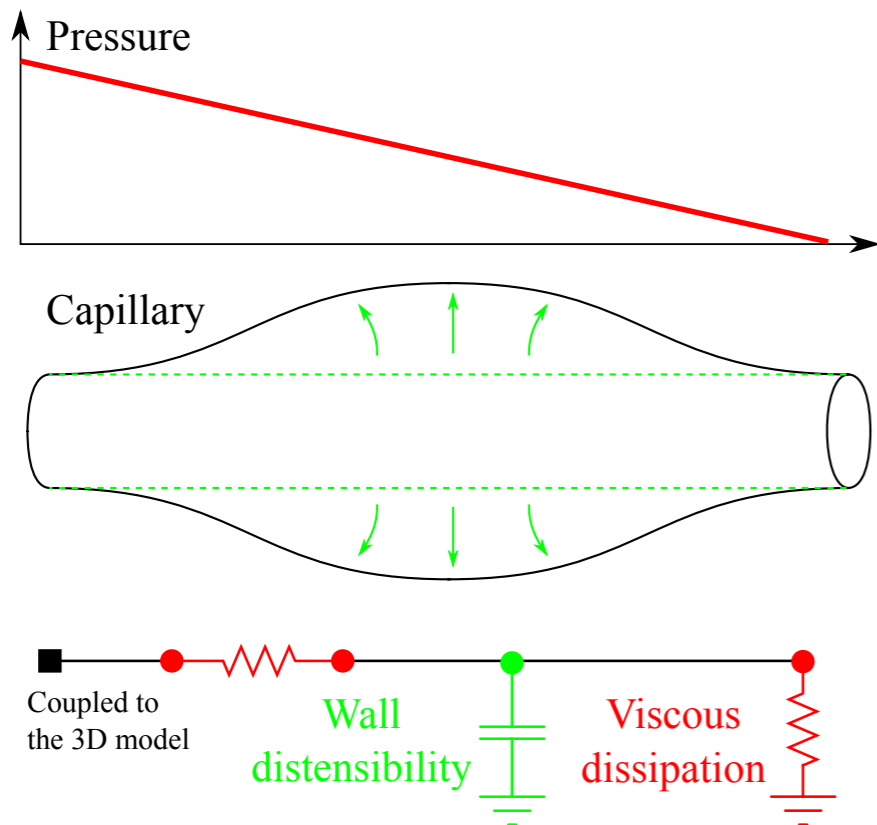
Cardiovascular Model Fidelity



0D Lumped Parameter Circuits

Circuit Analogy:

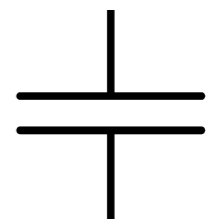
Flow \longleftrightarrow Current
 Pressure Drop \longleftrightarrow Voltage
 Ohm's Law $V=IR$



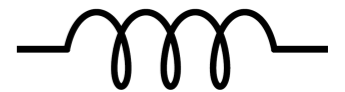
$$\Delta P = RQ$$



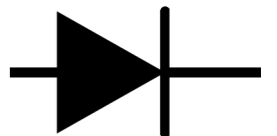
$$Q = C \frac{dP}{dt}$$



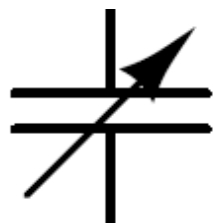
$$\Delta P = L \frac{dQ}{dt}$$



$$Q = \frac{|Q| + Q}{2}$$



$$P = A(t)E(V - V_0) + P_0 \left(e^{K(V-V_0)} - 1 \right)$$

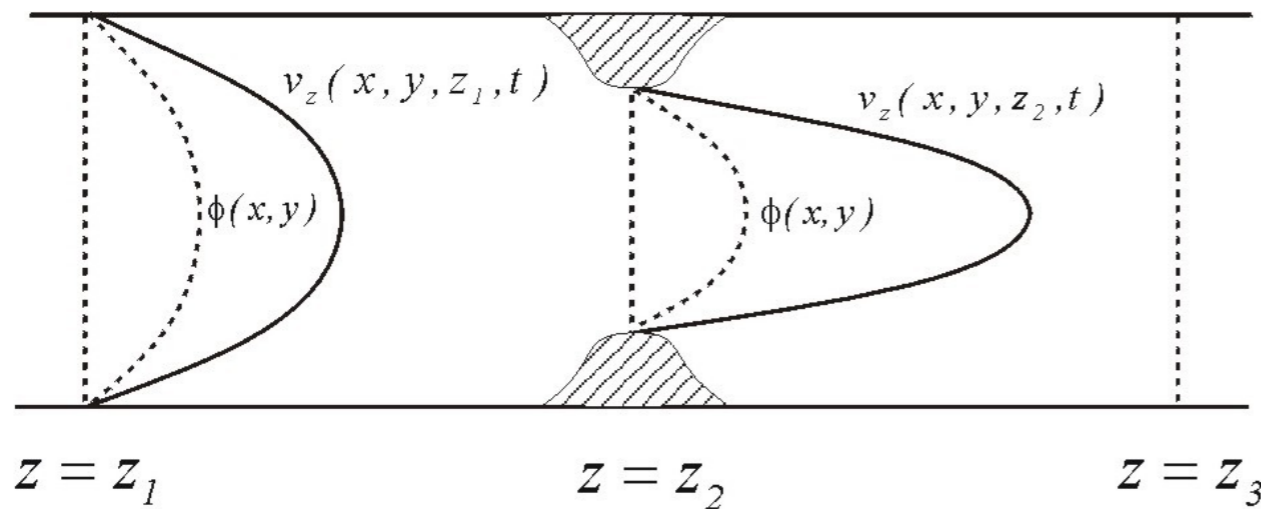


1D Wave Propagation Models

NS Eqs: Conservation of mass and momentum

$$\frac{\partial u_z}{\partial z} + \frac{1}{r} \frac{\partial}{\partial r} (r u_r) + \frac{1}{r} \frac{\partial}{\partial \theta} (u_\theta) = 0$$

$$\frac{\partial u_z}{\partial t} + u_z \frac{\partial u_z}{\partial z} + u_r \frac{\partial u_z}{\partial r} + \frac{u_\theta}{r} \frac{\partial u_z}{\partial \theta} + \frac{1}{\rho} \frac{\partial p}{\partial z} = \nu \frac{\partial}{\partial r} \left(r \frac{\partial u_z}{\partial r} \right) + \nu \frac{\partial^2 u_z}{\partial z^2} + \frac{\nu}{r^2} \frac{\partial^2 u_z}{\partial \theta^2}$$



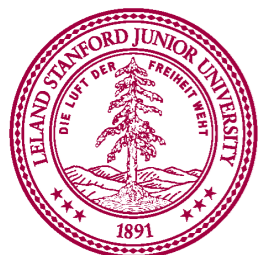
Integrate over vessel cross section
Assume parabolic velocity profile

$$\frac{\partial A}{\partial t} + \frac{\partial Q}{\partial z} = 0$$

$$\frac{\partial Q}{\partial z} + \frac{\partial}{\partial z} \left(\frac{4}{3} \frac{Q^2}{A} \right) + \frac{A}{\rho} \frac{\partial P}{\partial z} = -8\pi\nu \frac{Q}{A} + \nu \frac{\partial^2 Q}{\partial z^2}$$

1D PDE for flow rate and area

Hughes and Lubliner, *Mathematical Biosciences*, 1973

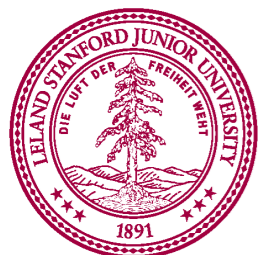
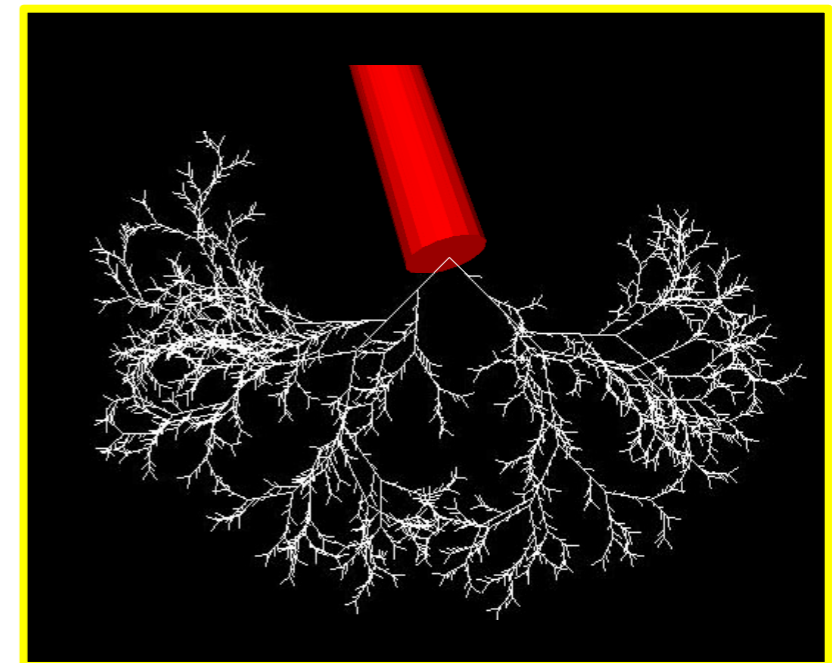
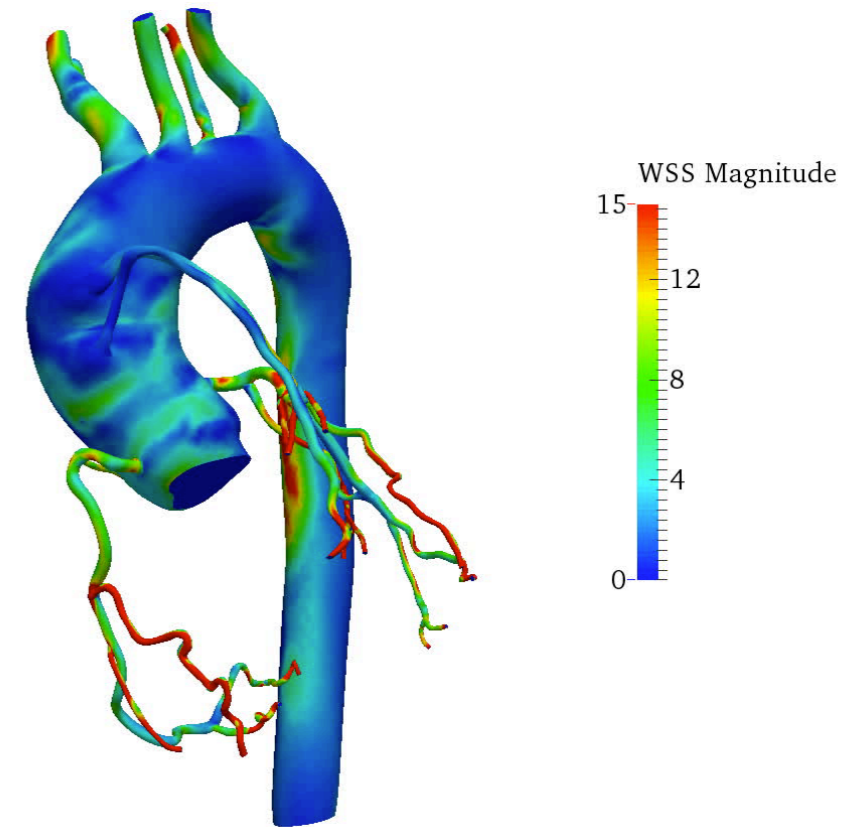


3D Computational Fluid Dynamics

$$\rho \vec{v}_{,t} + \rho \vec{v} \cdot \nabla \vec{v} = -\nabla p + \nabla \cdot \tau + \vec{f}$$
$$\nabla \cdot \vec{v} = 0$$

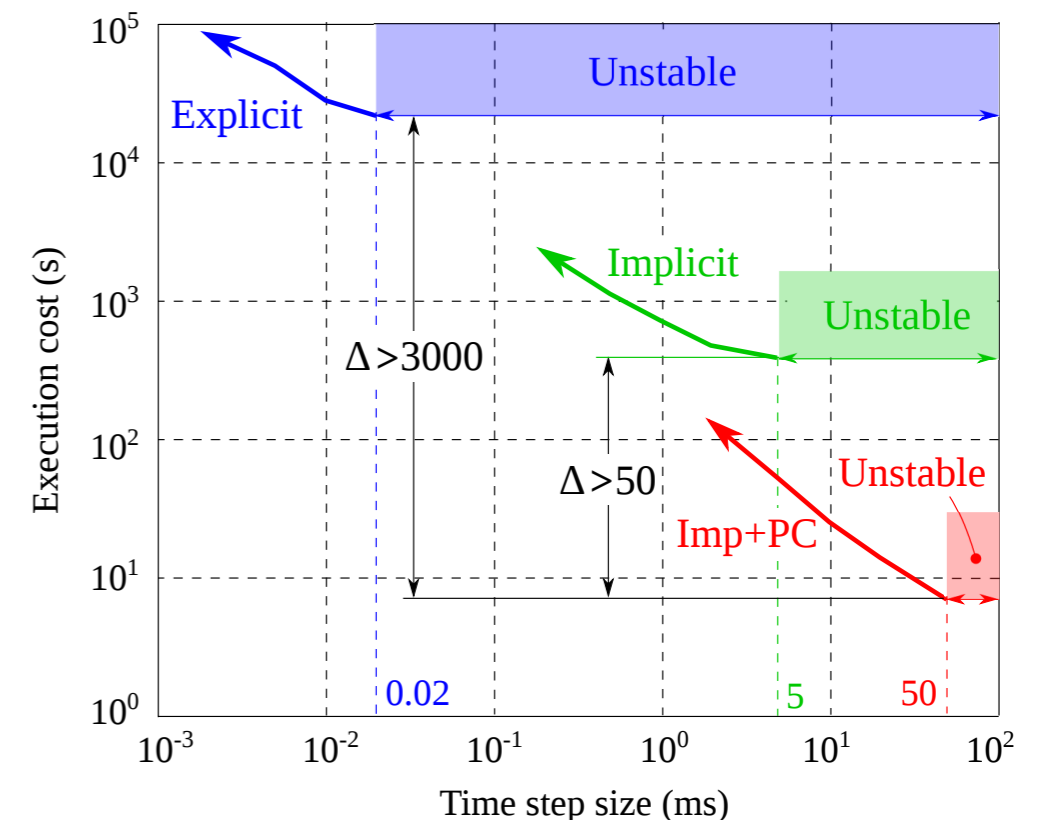
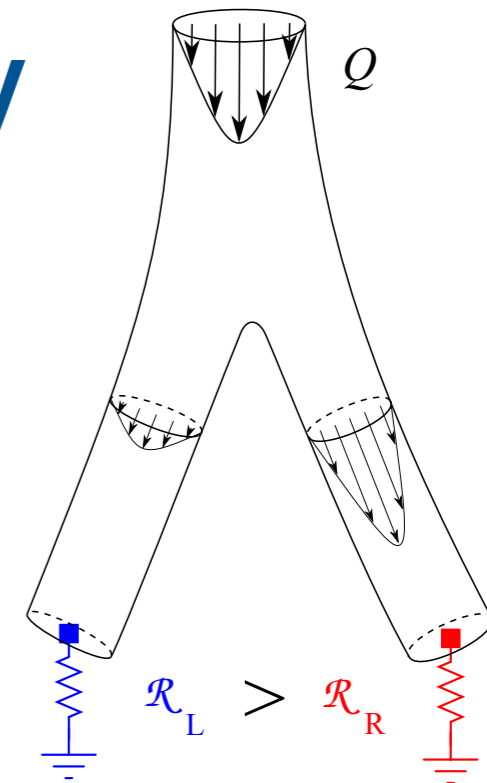
$$\begin{bmatrix} K & G \\ D & L \end{bmatrix} \begin{bmatrix} y_u \\ y_p \end{bmatrix} = - \begin{bmatrix} R_m \\ R_c \end{bmatrix}$$

- Finite element method SUPG
- Generalized-alpha time discretization
- Linear tetrahedral elements
- Physiologic boundary conditions

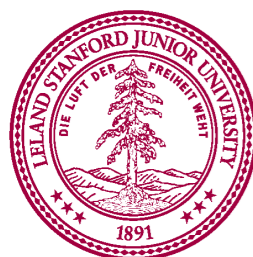


Solver Methodology

- Implicit coupling to LPN BCs
- Backflow stabilization
- Fluid structure interaction with
 - Coupled momentum method
 - ALE for large deformations
- Variable wall material properties: thickness, elastic modulus
- Custom linear solver with resistance-based preconditioner



Esmaily-Moghadam, Bazilevs, Marsden, Comp. Mech. 2013
 Marsden and Esmaily Moghadam, AMR, 2015
 Figueroa and Taylor, CMAME 2006

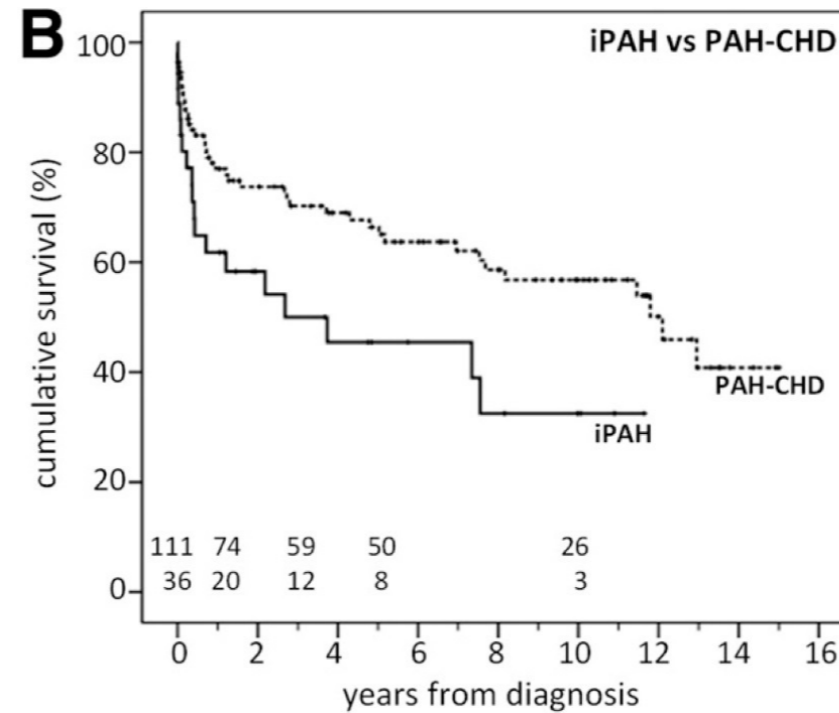
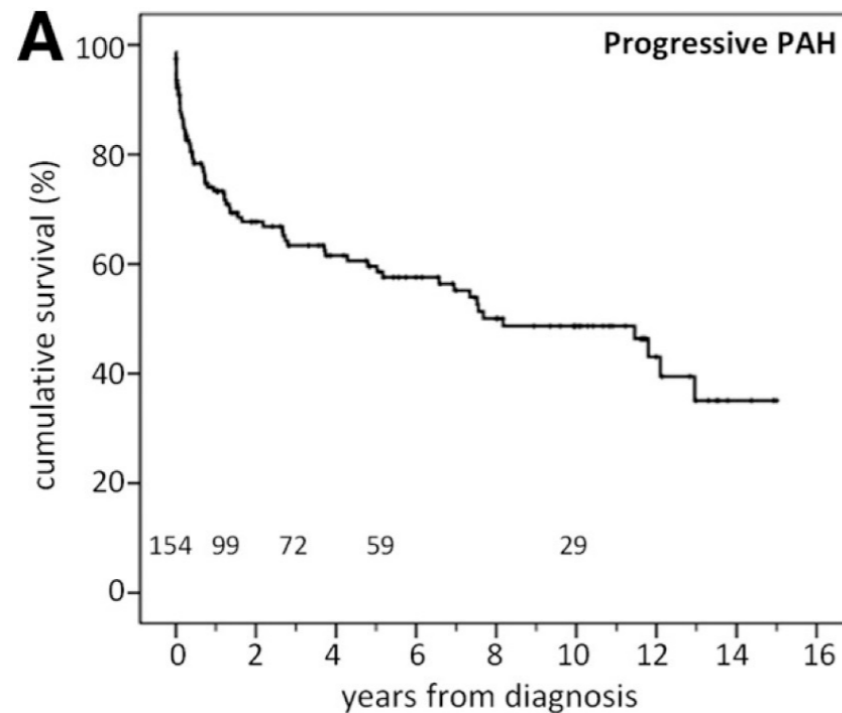


Clinical Examples

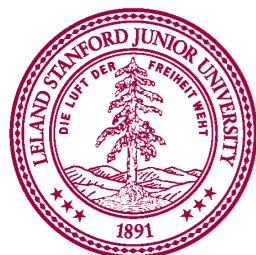
Pulmonary Hypertension

Pulmonary Hypertension

- Severely elevated pulmonary pressure $>25\text{mmHg}$ and vascular resistance $>3\text{WU}$
- Progression is highly variable and poorly understood (**5-year survival rates: 60-70%**).
- Need predictive tools for clinical decision making
 - When to list patients for transplant?

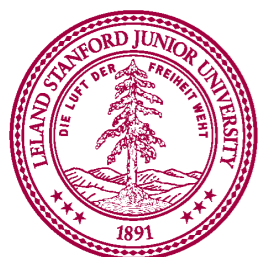
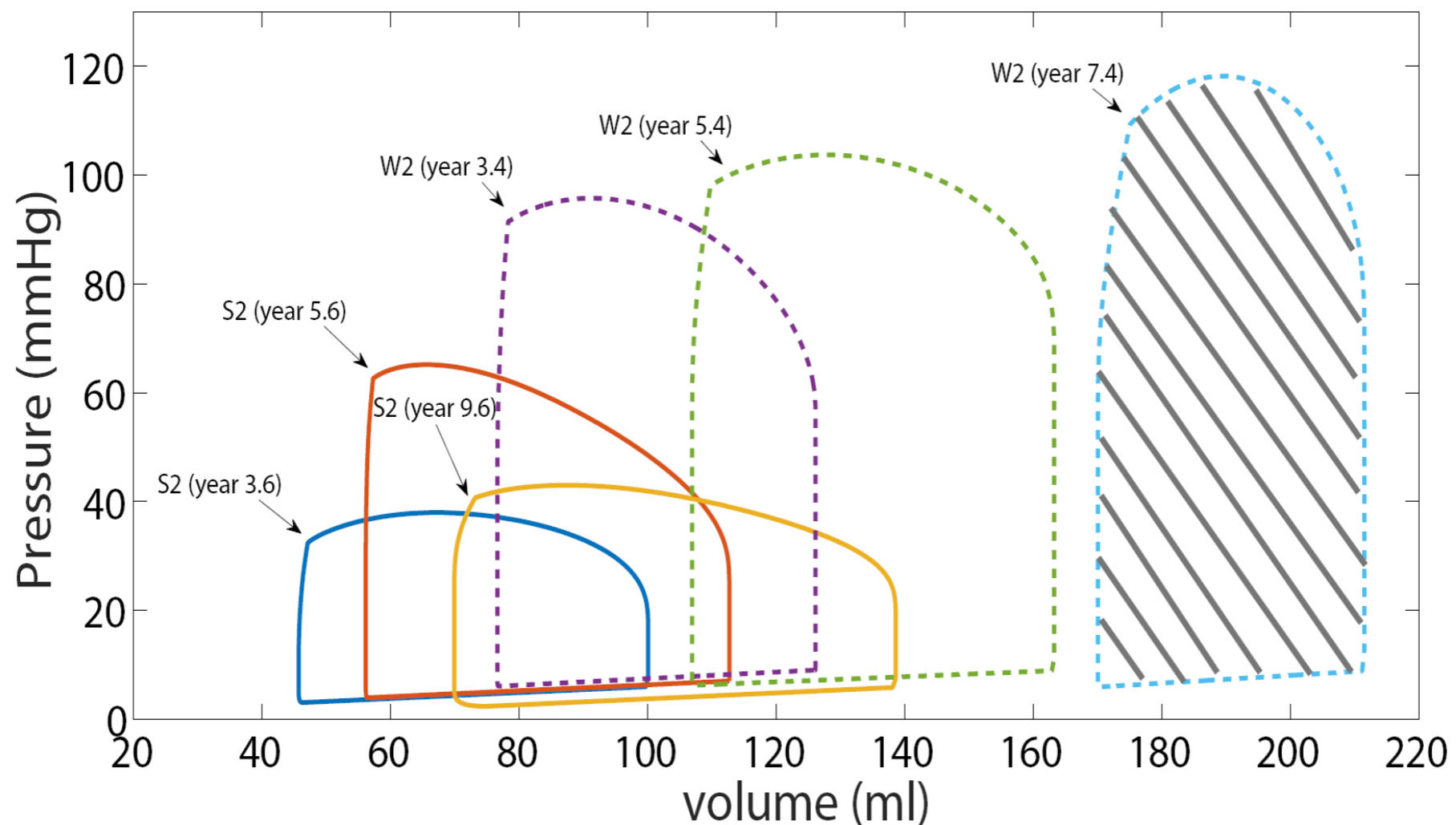


Van Loon et al., Circulation, 2011



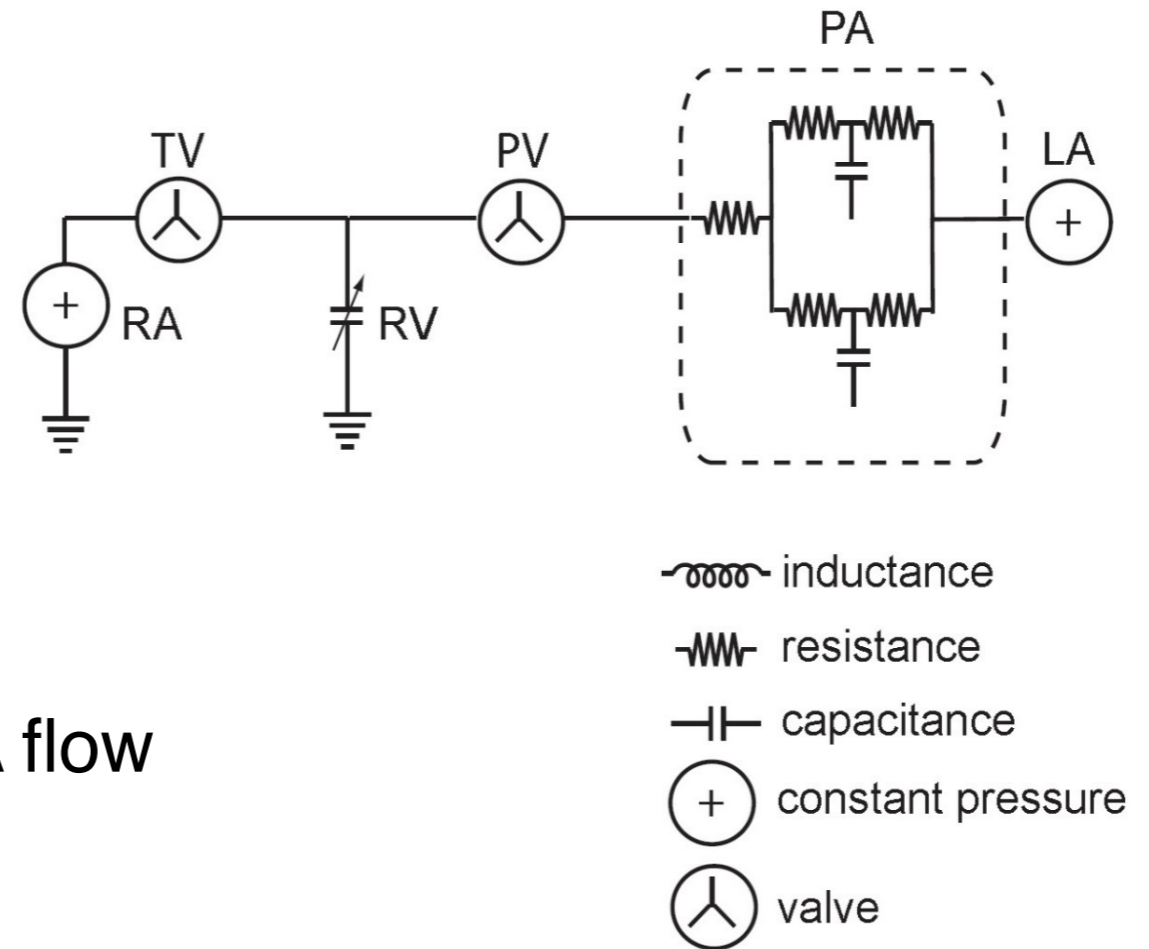
Right Ventricular Stroke Work

- Stroke work found from area enclosed in cardiac pressure-volume loop
 - Work done by the heart in each beat
- Challenging to measure routinely in clinic
- Use LPN heart model to compute RVSW from clinical data



Tuning to clinical data

- Right heart catheterization (RHC):
 - RV and PA pressures
- Magnetic resonance imaging (MRI):
 - RV volumes (EDV and ESV) and PA flow
- Lumped parameter network (LPN): use electric circuits to model hemodynamics
- Tune LPN to match clinical data with optimization



Study Design: Inclusion Criteria

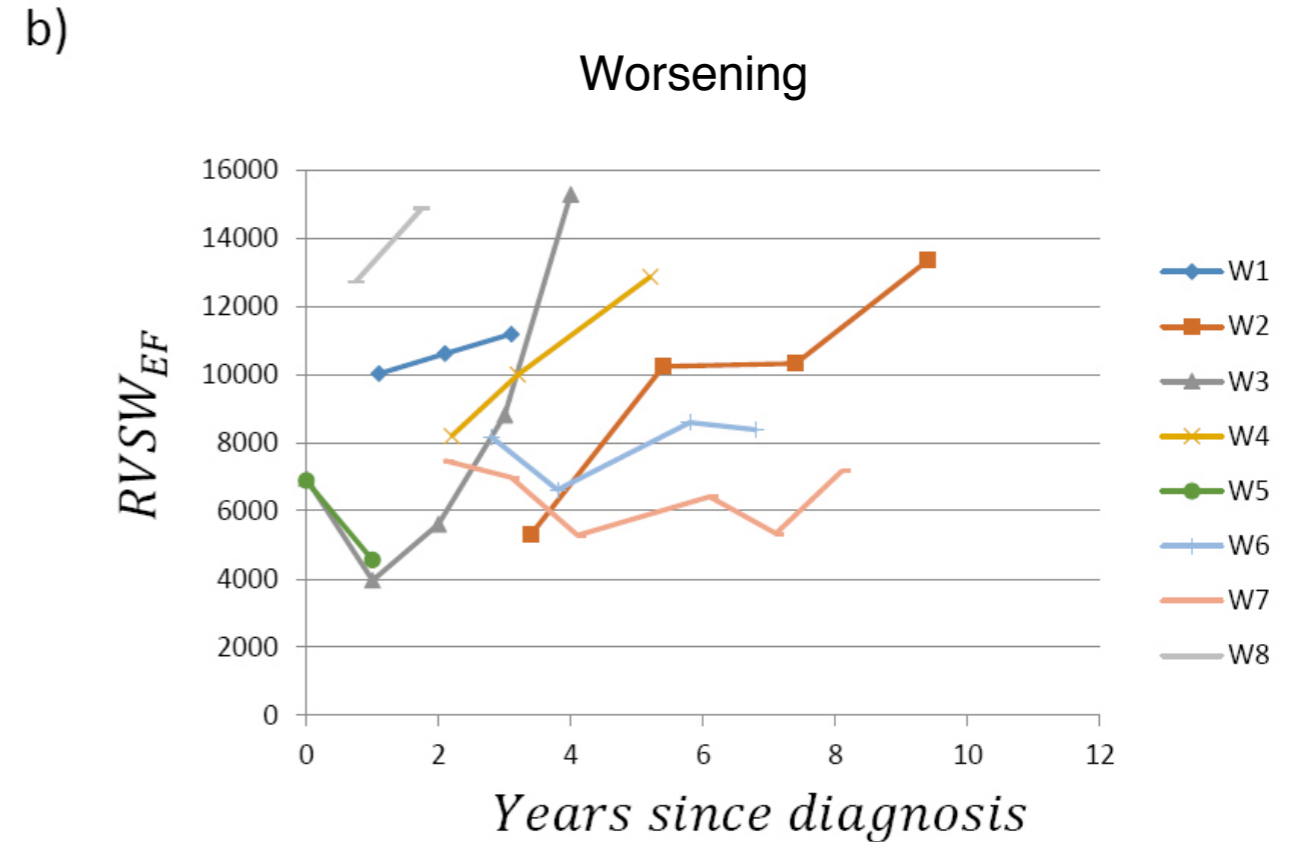
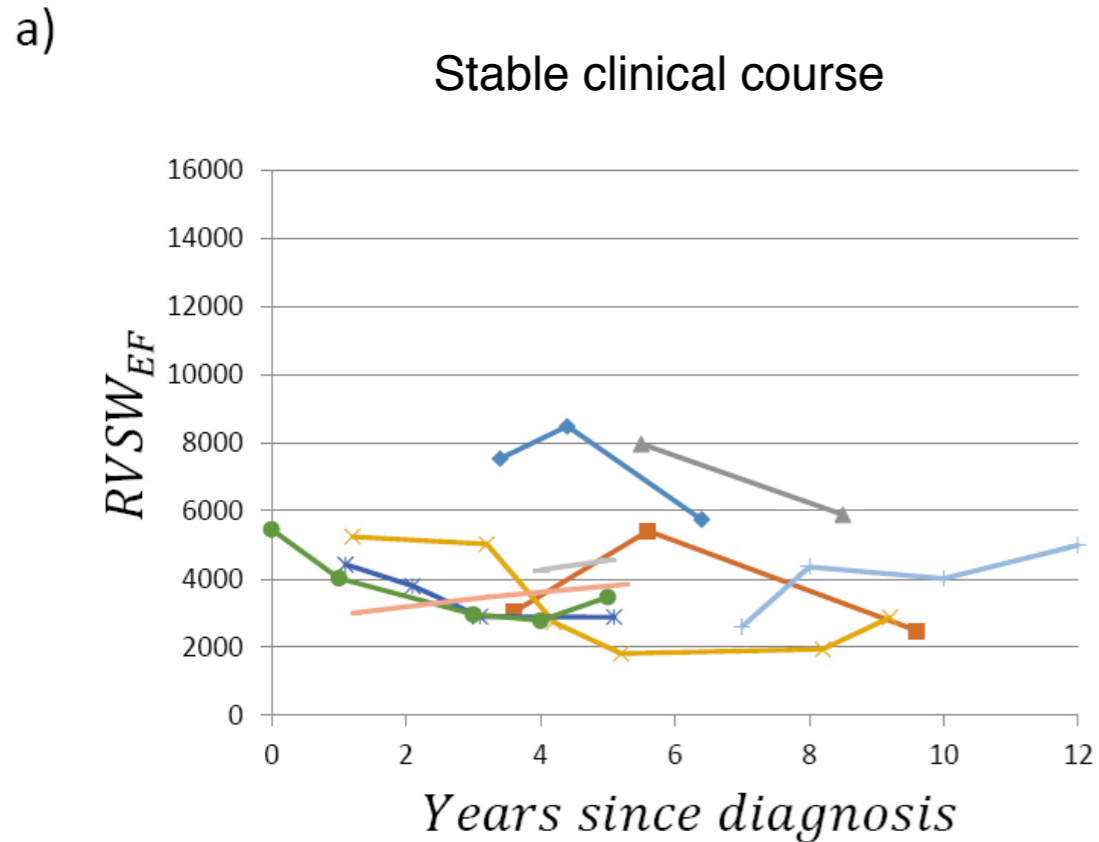
	Stable n=9	Worsening* n=8
Male	n=4	n=4
IPAH	n=5	n=7
Follow-up:	4.2 (1.2-8)	3.7 (1.1-6)

- Inclusion criteria
 - Age < 18
 - IPAH or PAH-CHD
 - multiple paired RHC/MRI ($n \geq 2$)
 - 17 patients with 61 data points

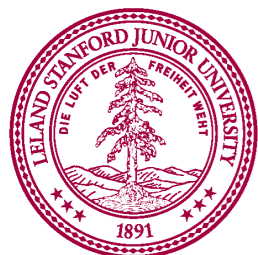
Clinical worsening: death, listed or considered for transplantation, poor hemodynamic responses to maximal therapy



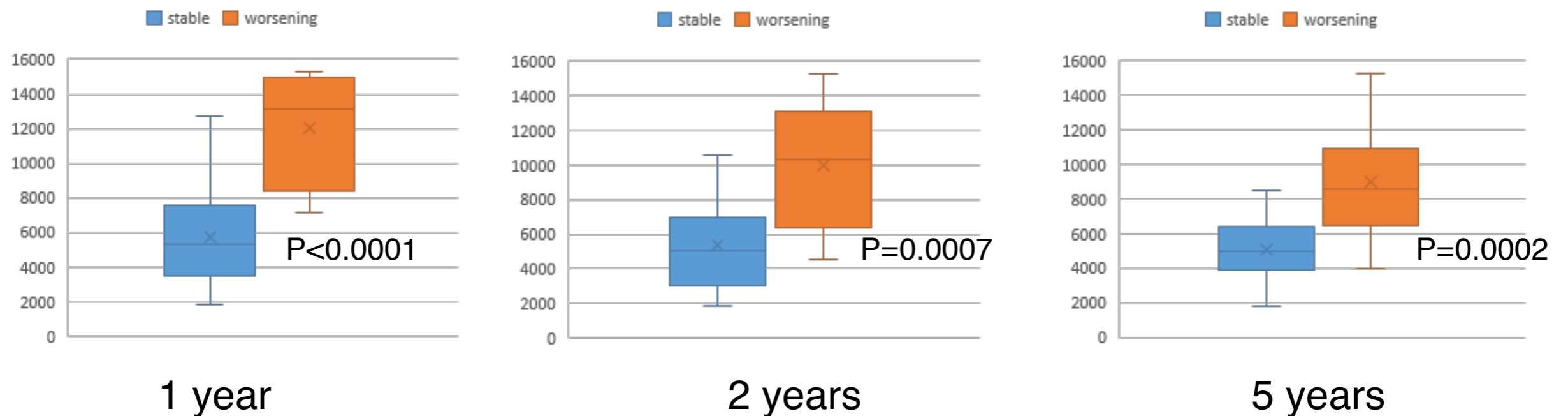
Stroke work and disease progression



Patients with clinical worsening have increased RVSWEF.

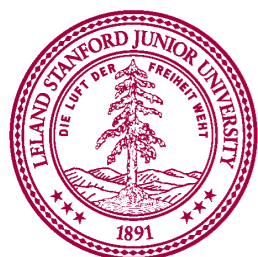


RV Stroke Work predicts Clinical Worsening in PH



RVSW outperformed other standard clinical metrics used to predict the need for heart transplant

Yang, W., Marsden, A.L., Ogawa, M.T., Sakarovitch, C., Phillips, K. K., Rabinovitch, M., Feinstein, J.A., "Right Ventricular Stroke Work Correlates with Outcomes in Pediatric Pulmonary Arterial Hypertension," submitted for review.

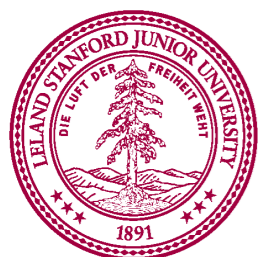
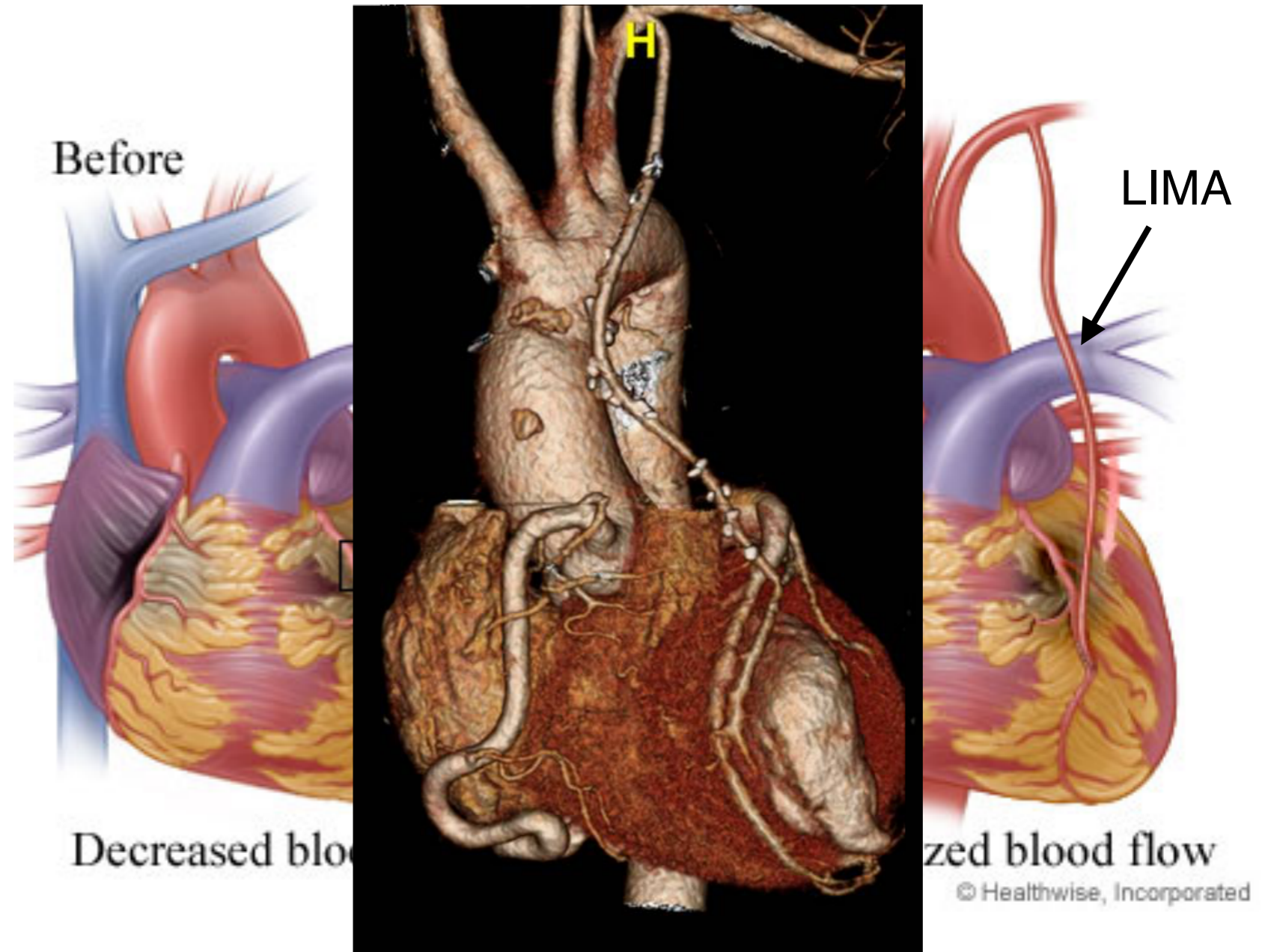


Clinical Examples

Coronary Artery Bypass Surgery

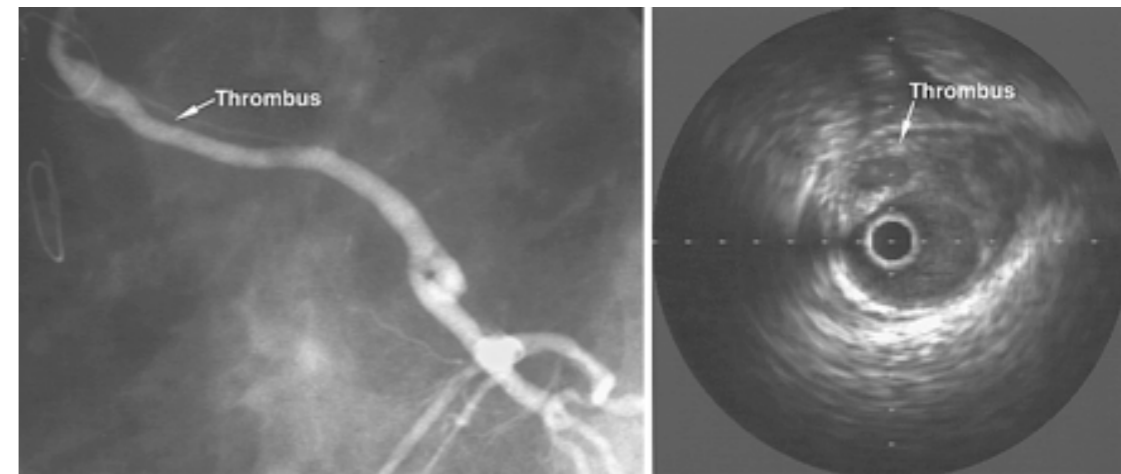
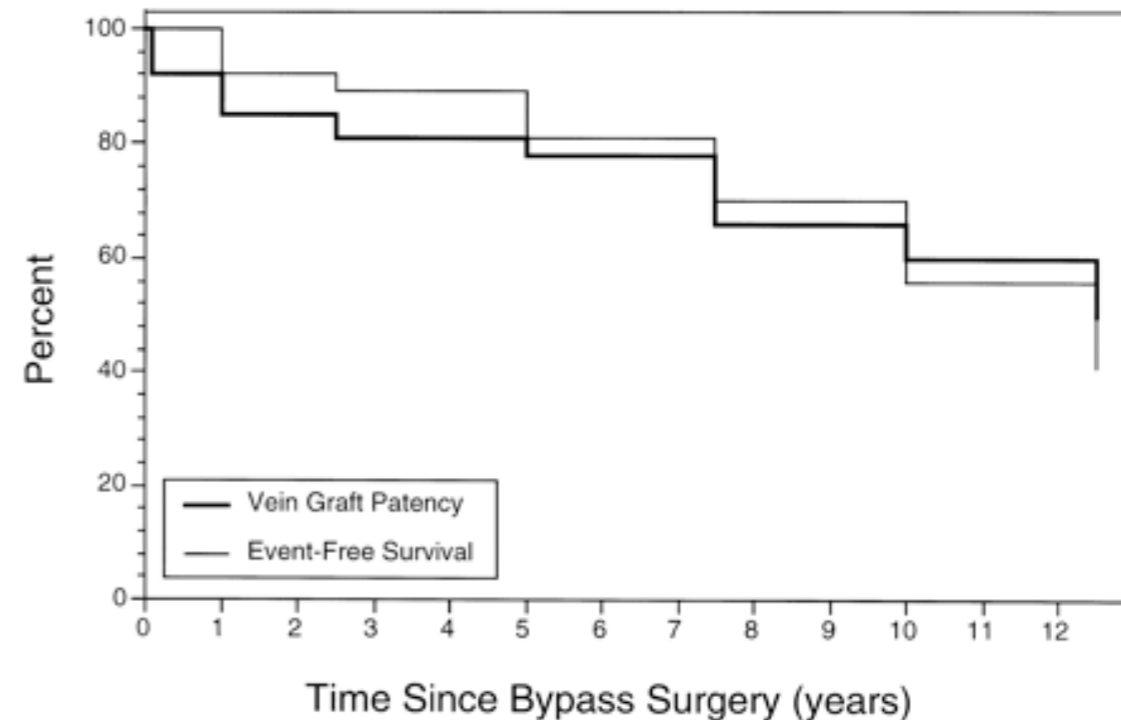
CABG Surgery

- CABG surgery performed in ~400,000 cases annually in US
- Graft options: arterial graft (LIMA), saphenous vein graft (SVG), artificial grafts
- Most patients require multiple grafts
- SVGs are used in majority of patients (70%)

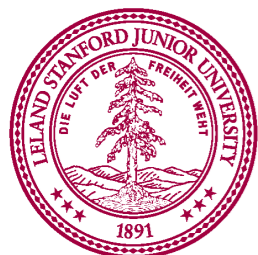


CABG: Vein Graft Failure

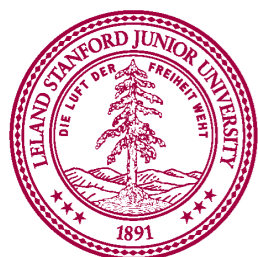
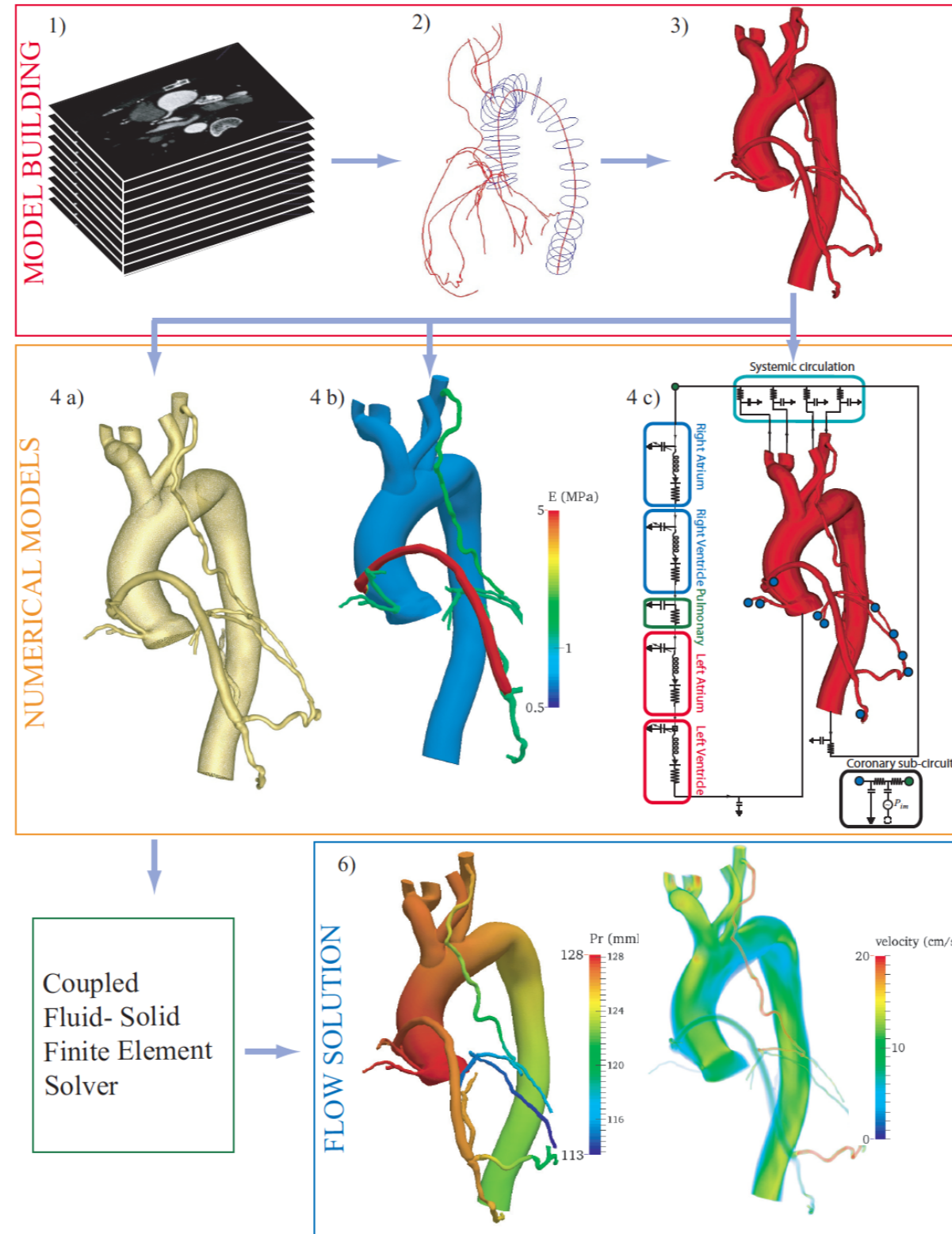
- Vein graft failure is a significant clinical problem
 - 5-10% of SVGs fail within 1 month
 - 40-50% fail within 10 years
- Clamps removed: vein subjected to 20X increase in pressure, 4X increase in flow
- How does vessel adapt to changing mechanical loads?
- *What hemodynamic and biomechanical conditions lead to vein graft failure?*



Aortocoronary Saphenous Vein Graft Disease: Pathogenesis, Predisposition, and Prevention, Joseph G. Motwani, MD; Eric J. Topol, MD, Circulation 1998

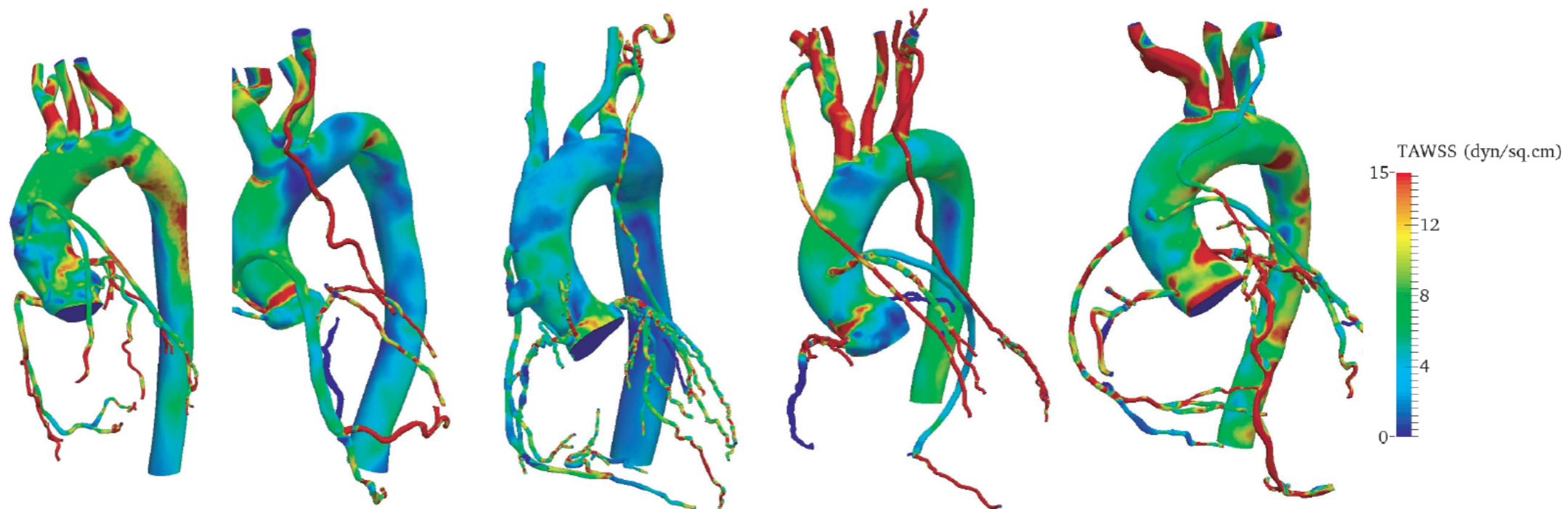


CABG Simulation Pipeline

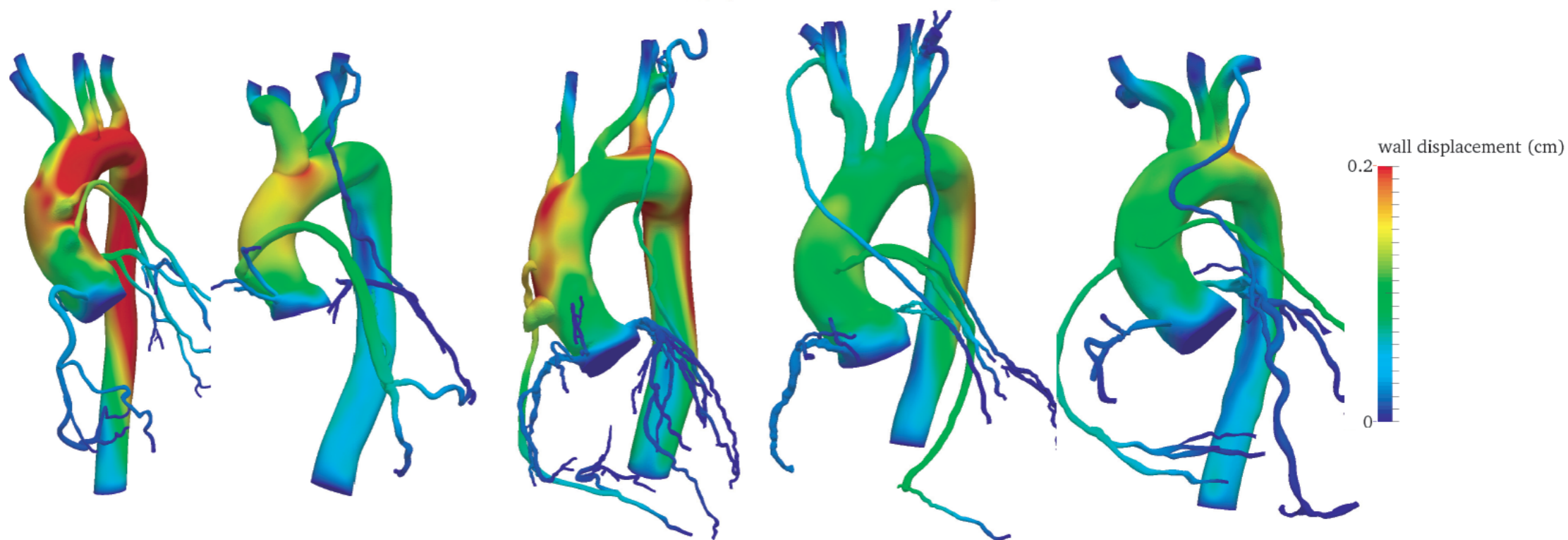


Simulation Results

WSS



Wall Strain



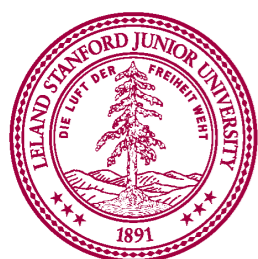
Patient 1

Patient 2

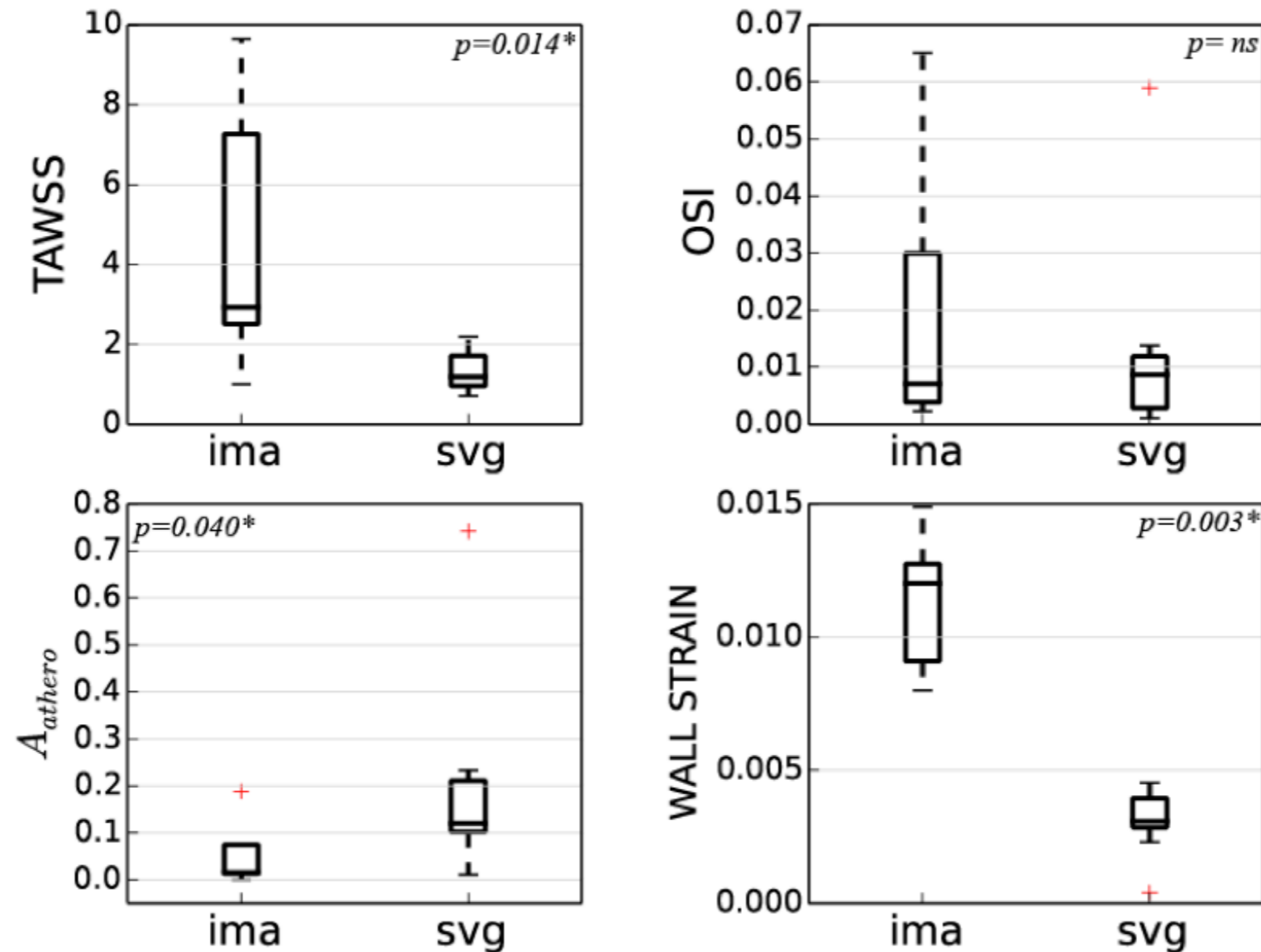
Patient 3

Patient 4

Patient 5

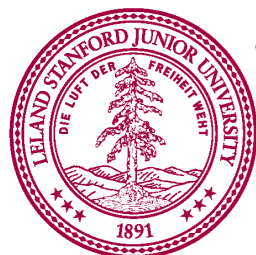


Mechanical Stimuli: Arterial vs. Vein Grafts



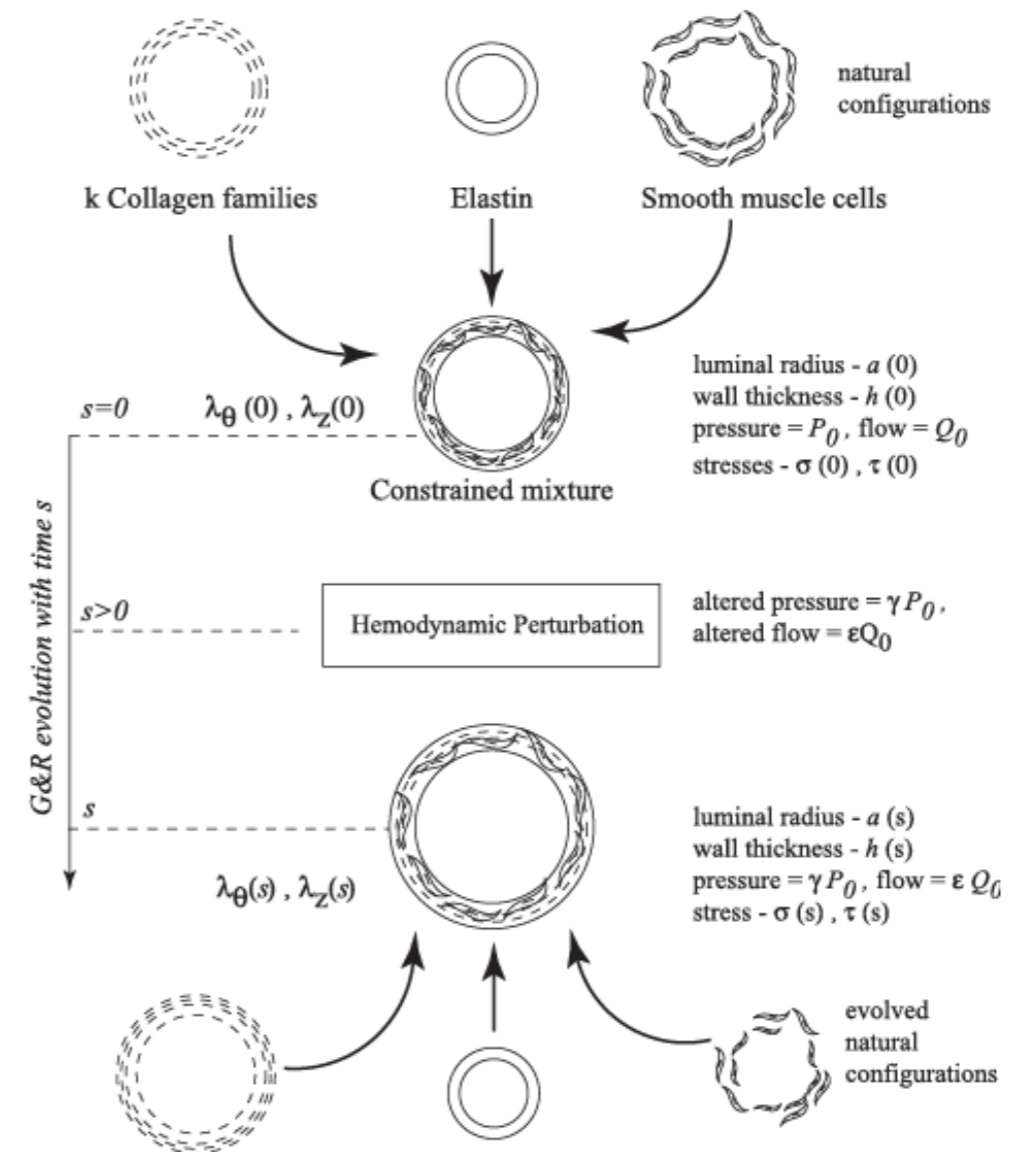
Statistically significant differences in WSS, area of low WSS, wall strain

Ramachandra, A. B., Kahn, A. M., Marsden, A.L., "Patient specific simulations reveal significant differences in mechanical stimuli in venous and arterial coronary grafts," *Journal of Cardiovascular Translational Research*, Vol. 9 (4), pp 279–290, (2016).

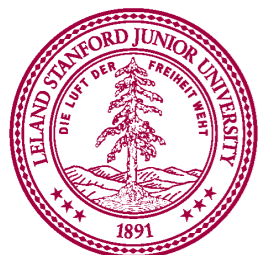


Growth and Remodeling

- Adapted Humphrey arterial G&R model to veins
- Predict response to changes in hemodynamics (pressure, shear stress)
- radius, thickness, wall composition
- Test hypotheses of vein graft failure



What is the biomechanical response to altered hemodynamics and wall mechanics in a vein graft?



G&R Model

Equilibrium Equations

$$P(s)a(s) = T_\theta(s) = \lambda_\theta \sum \frac{\partial W^k}{\partial \lambda_\theta} \quad \text{Circumferential}$$

$$\frac{f(s)}{\pi(2a(s) + h(s))} = T_z(s) = \lambda_z \sum \frac{\partial W^k}{\partial \lambda_z} \quad \text{Axial}$$

Constitutive Equations

$$W^{elastin} = \frac{c^e}{2} (\lambda_\theta^2 + \lambda_z^2 + \lambda_r^2 - 3) \quad \text{Elastin}$$

$$W^k = \frac{c_1^k}{4c_2^k} (e^{(c_2^k(\lambda)^2 - 1)^2} - 1) \quad \text{SMC + Collagen}$$

Mass Addition and Removal

$$m^k = m_0^k (1 + K_1^k \Delta \sigma - K_2^k \Delta \tau_w) \quad \text{Addition}$$

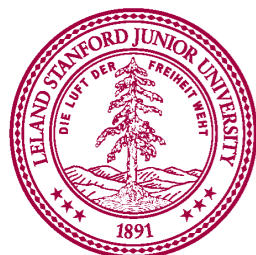
$$q^k(s, \tau) = e^{-\int_\tau^s K^k(\tilde{\tau}) d\tilde{\tau}} \quad \text{Removal}$$

$$K^k(\tilde{\tau}) = K_h^k + K_h^k \Delta \zeta(\tilde{\tau})^2$$

Evolution

$$M^k(s) = M^k(0)Q^k(s) + \int_0^s m^k(\tau)q^k(s, \tau)d\tau \quad \text{Mass}$$

$$W^k(s) = \frac{M^k(0)}{\rho(s)} Q^k(s) \widehat{W}^k(\mathbf{C}_{n(0)}^k(s)) + \int_0^s \frac{m^k(\tau)}{\rho(s)} q^k(s, \tau) \widehat{W}^k(\mathbf{C}_{n(\tau)}^k(s)) d\tau, \quad \text{Stored Energy}$$



Could gradual loading ameliorate vein maladaptation?

$$\mathcal{J}_{adapt} = \sqrt{\left(\frac{\tau_w^h - \tau_w}{\tau_w}\right)^2 + \left(\frac{\sigma_\theta^h - \sigma_\theta}{\sigma_\theta}\right)^2}$$

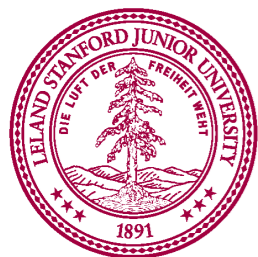
Measure deviation from homeostasis

Step Change

Gradual Change

Ramachandra, A. B., Sankaran, S., Humphrey, J.D., Marsden, A.L., “Computational simulation of the adaptive capacity of vein grafts in response to increased pressure,” *Journal of Biomechanical Engineering*, Vol. 137, pp. 031009-1, (2015).

Ramachandra, A. B., Humphrey, J. D., Marsden, A. L., “Gradual loading ameliorates maladaptation in computational simulations of vein graft growth and remodeling,” *Journal of the Royal Society Interface*, Vol. 14 (130), May 2017.

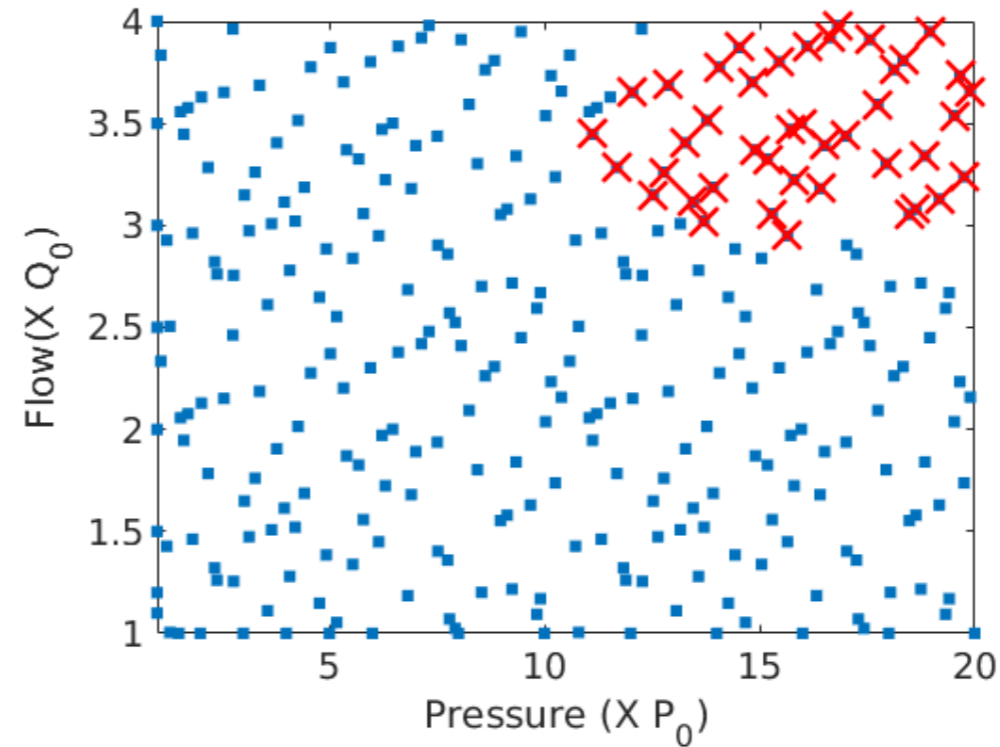
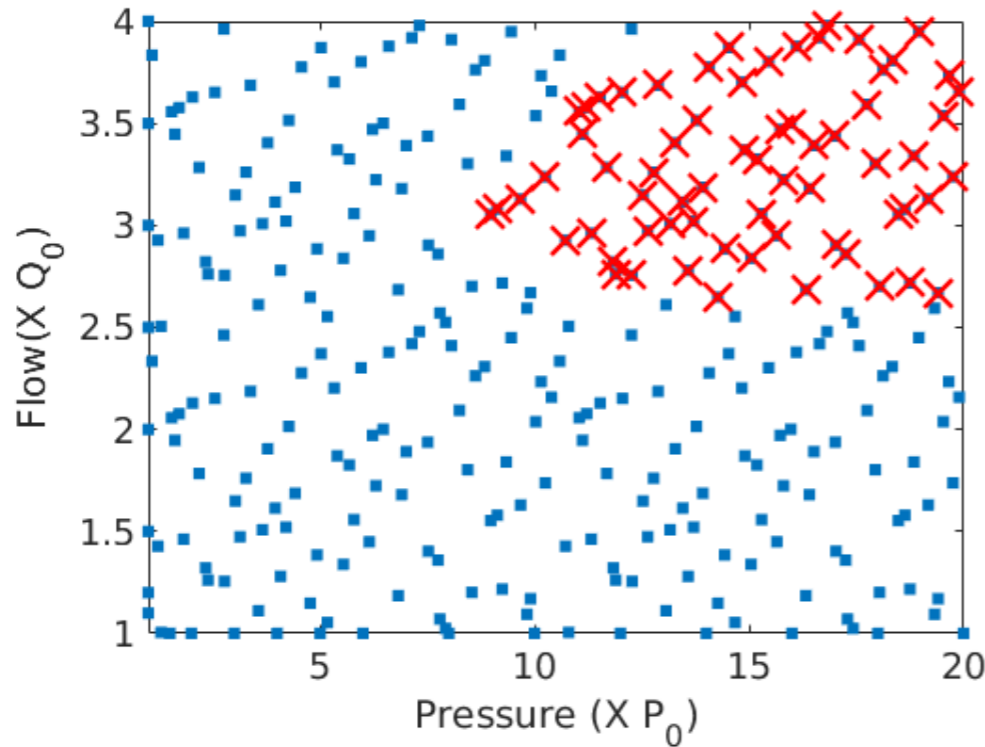


Step vs. Gradual Loading

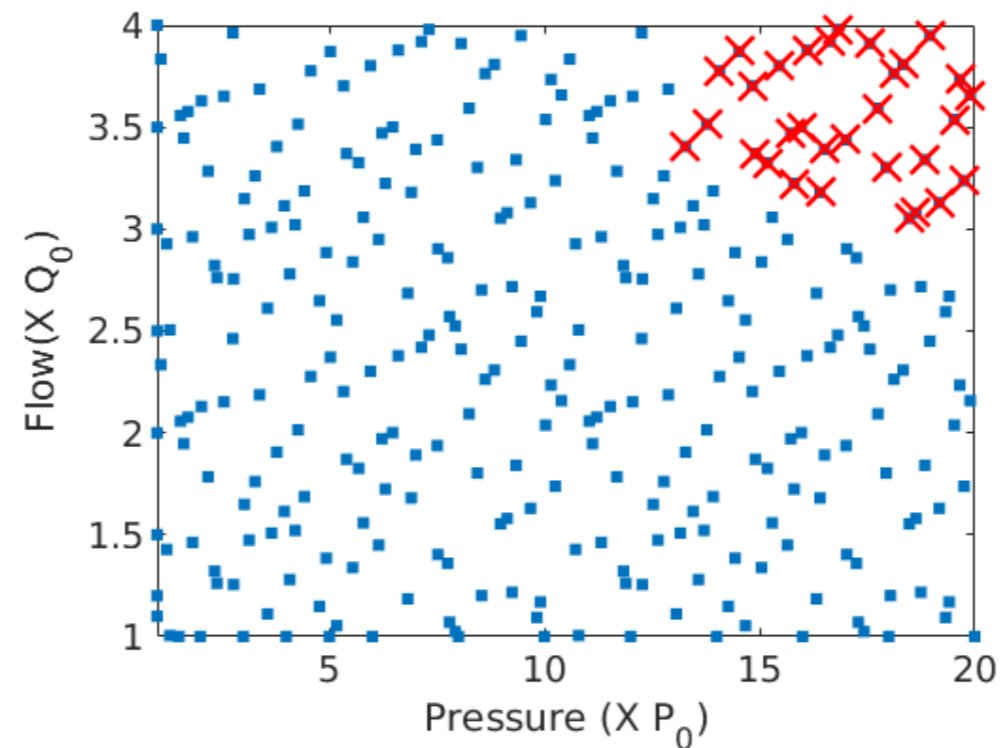
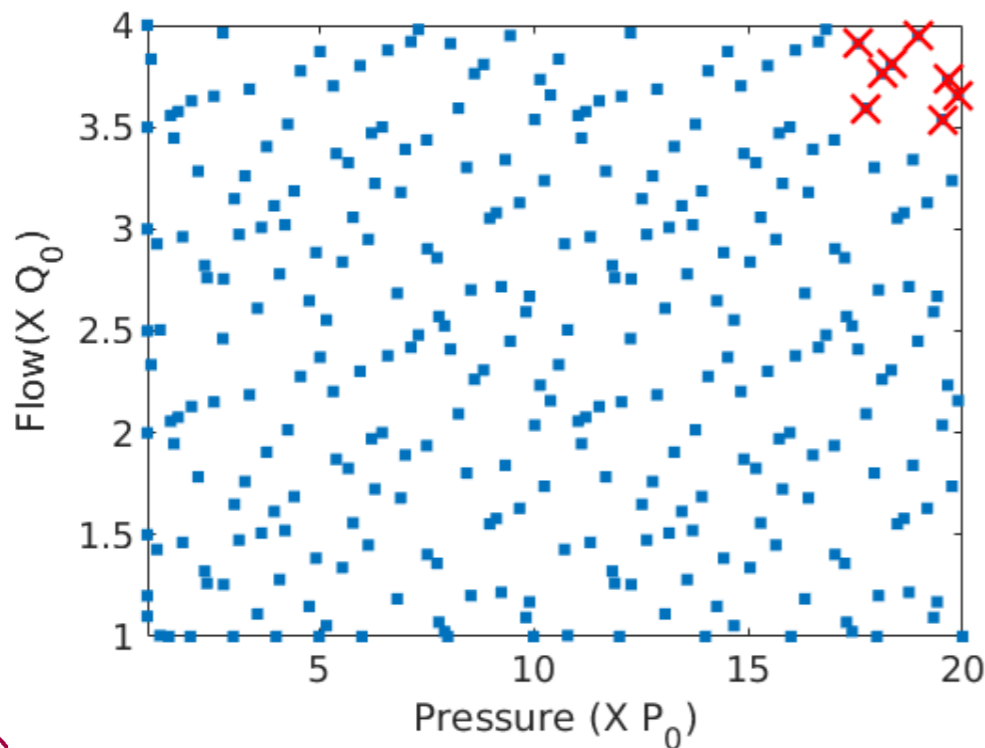
Step Pressure

Gradual Pressure – 3days

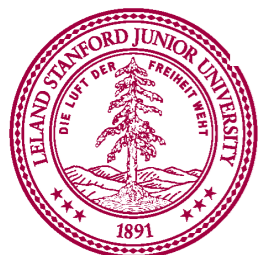
Step Flow



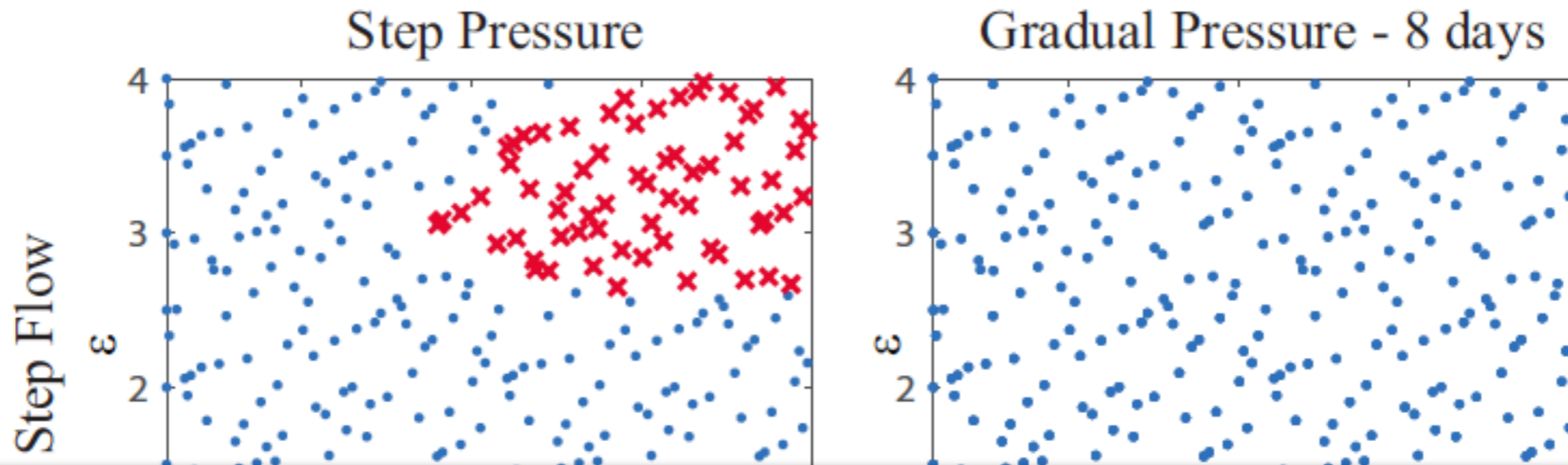
Gradual Flow- 3days



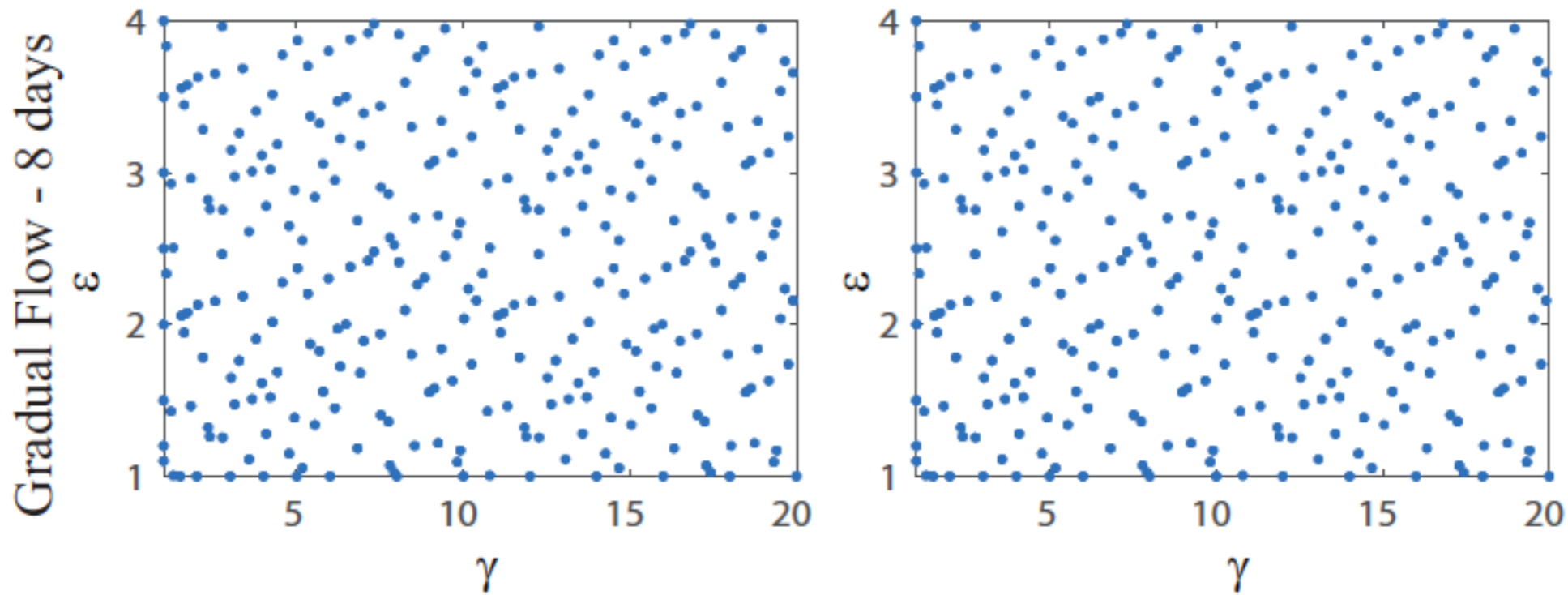
X = Model predicts failure
500 days post CABG



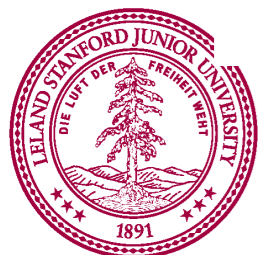
Step vs. Gradual Loading



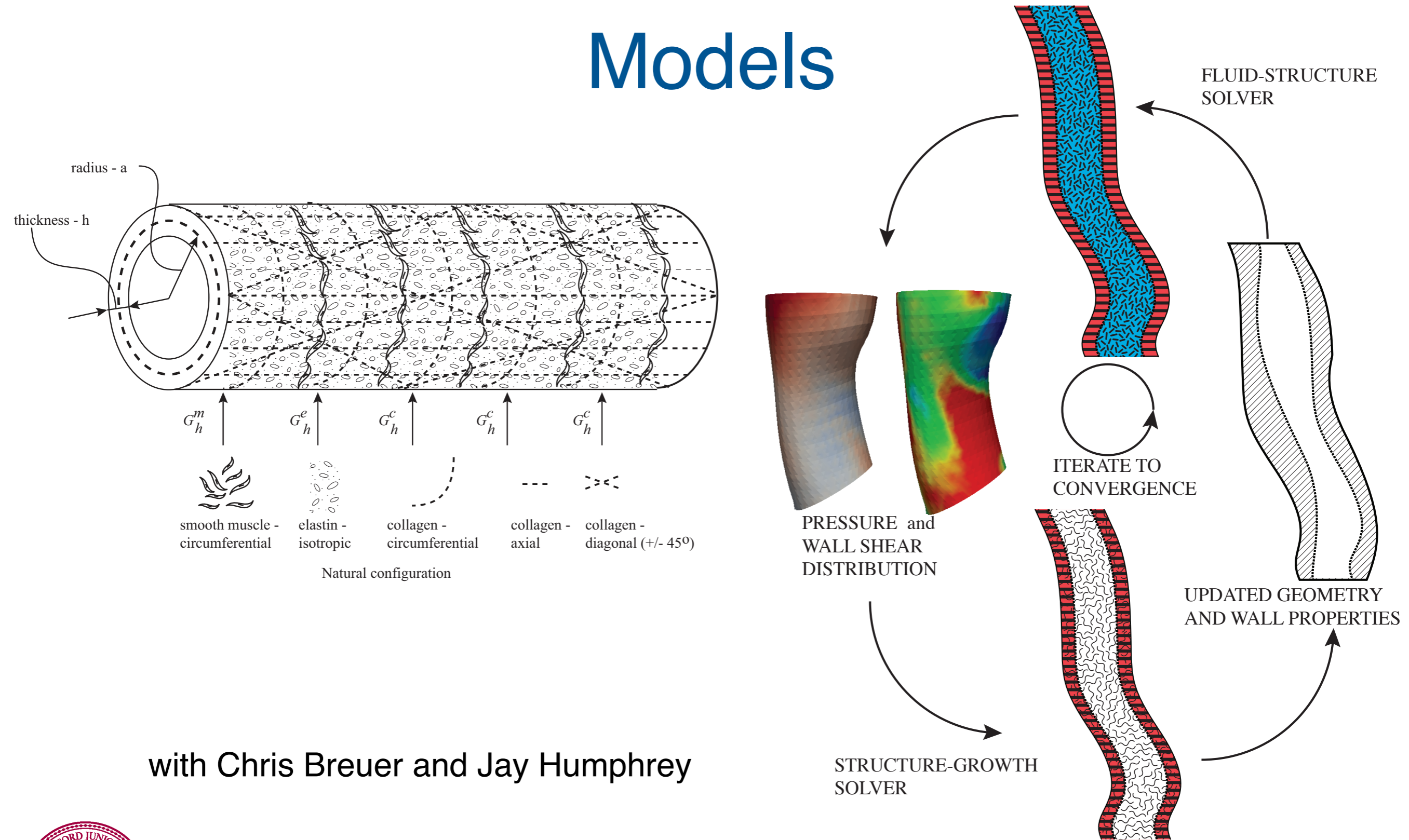
Applying gradual loading over the first 3-8 days ameliorates intermediate-term graft failure



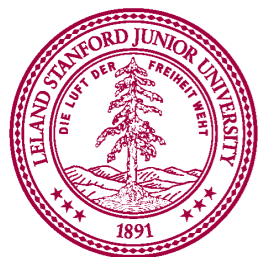
X = Model predicts failure
500 days post CABG



Towards Fluid-Solid-Growth Models

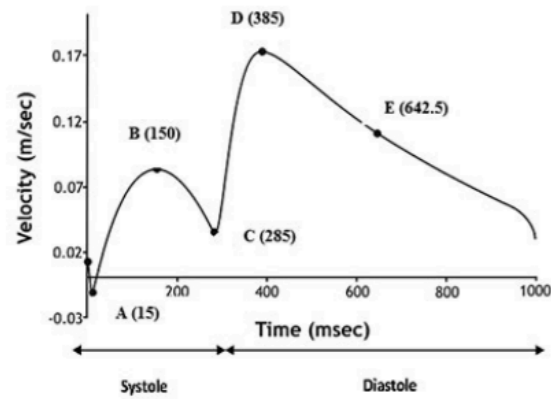


with Chris Breuer and Jay Humphrey

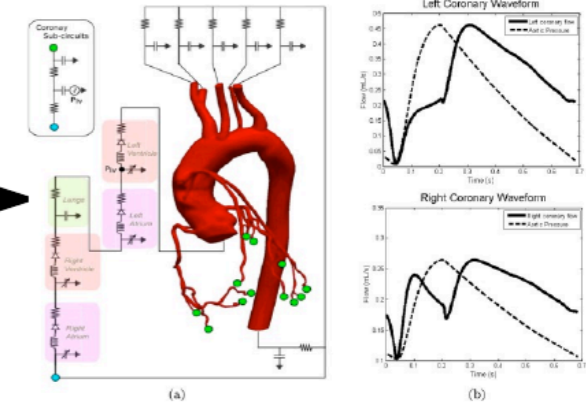


Parameter Estimation and UQ

With Daniele Schiavazzi, Notre Dame

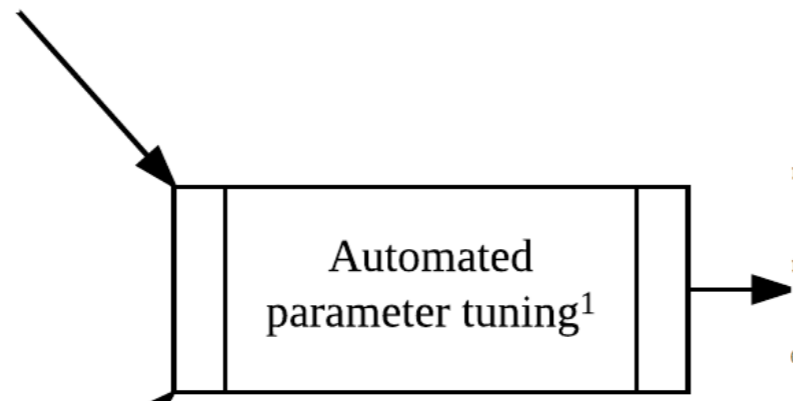


Patient-specific simulation

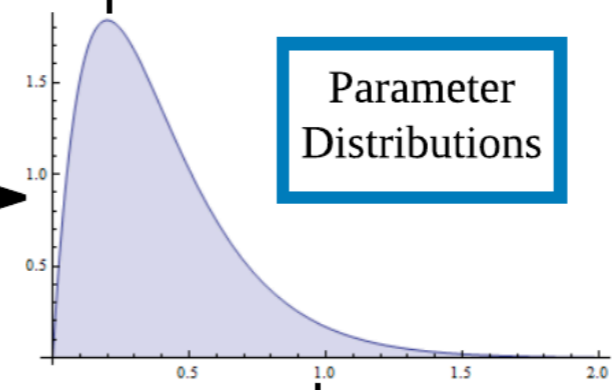


Literature data on coronary flow

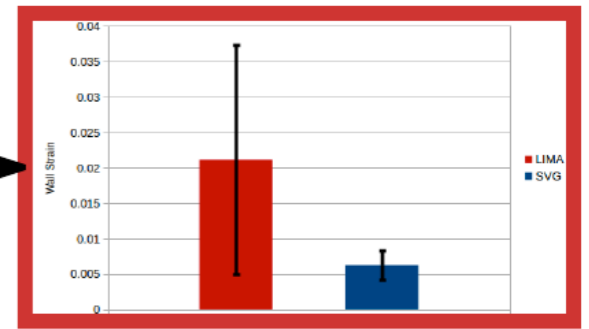
Patient Clinical Data
- Blood pressure
- Stroke Volume
- etc...



Optimal Parameter Values



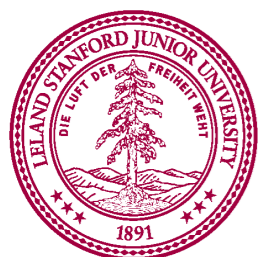
Uncertainty Quantification



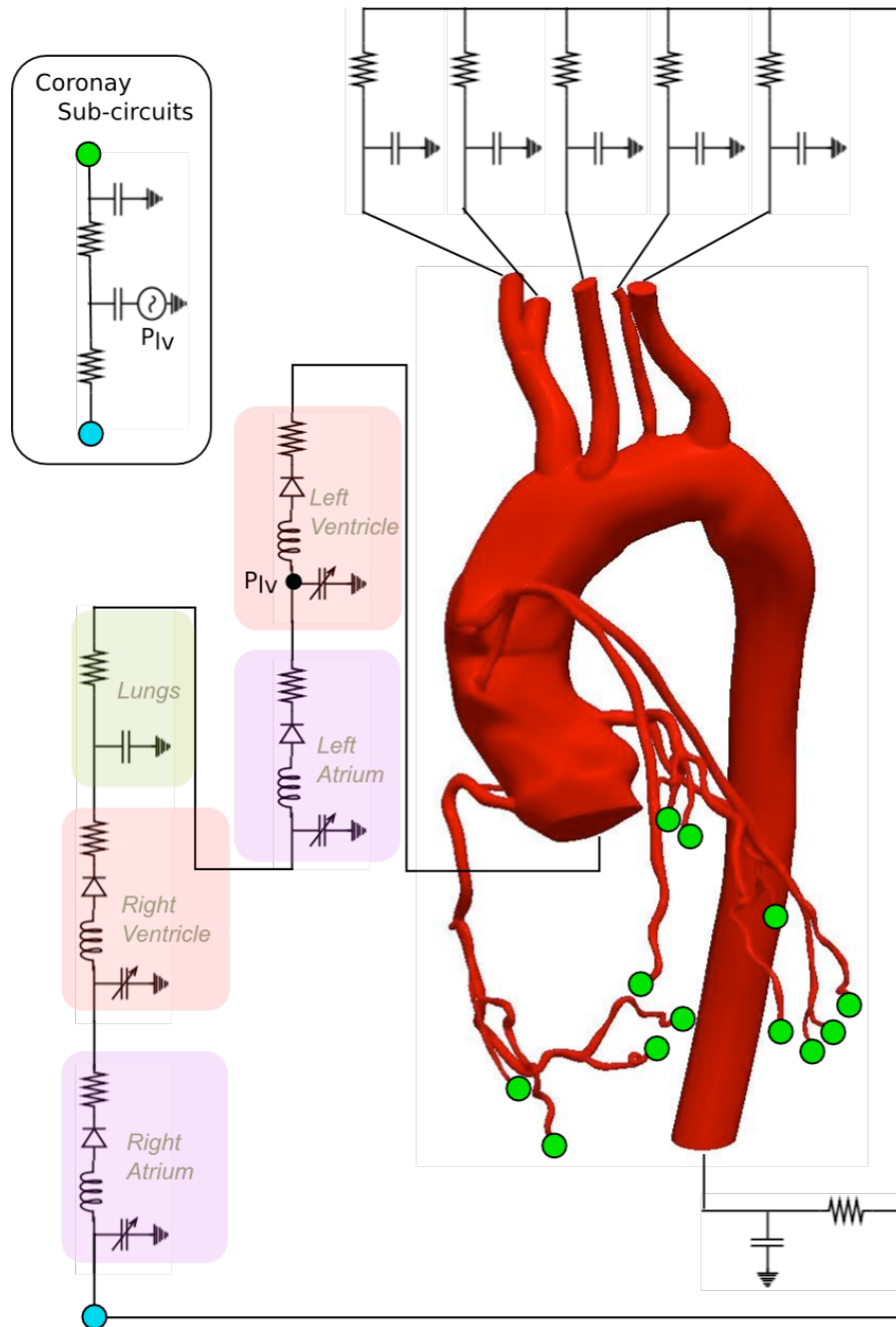
Output statistics

Schiavazzi, D. E., Doostan, A., Iaccarino, G., Marsden, A. L., "A Generalized Multi-resolution Expansion for Uncertainty Propagation with Application to Cardiovascular Modeling," *Computer Methods in Applied Mechanics and Engineering*, Vol. 314 (1), pp. 196-221, (2017).

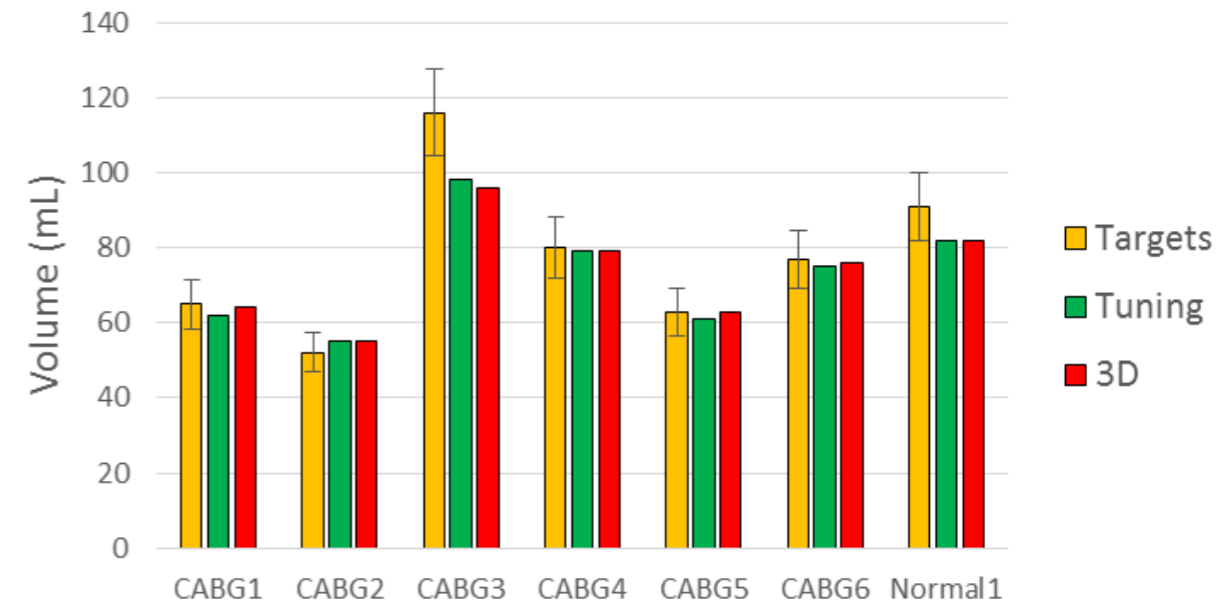
Schiavazzi, D. E., Hsia, T. Y., Marsden, A. L. "On a sparse pressure-flow rate condensation of rigid circulation models," *Journal of Biomechanics*, Vol. 49 (11), pp. 2174-2186, (2016).



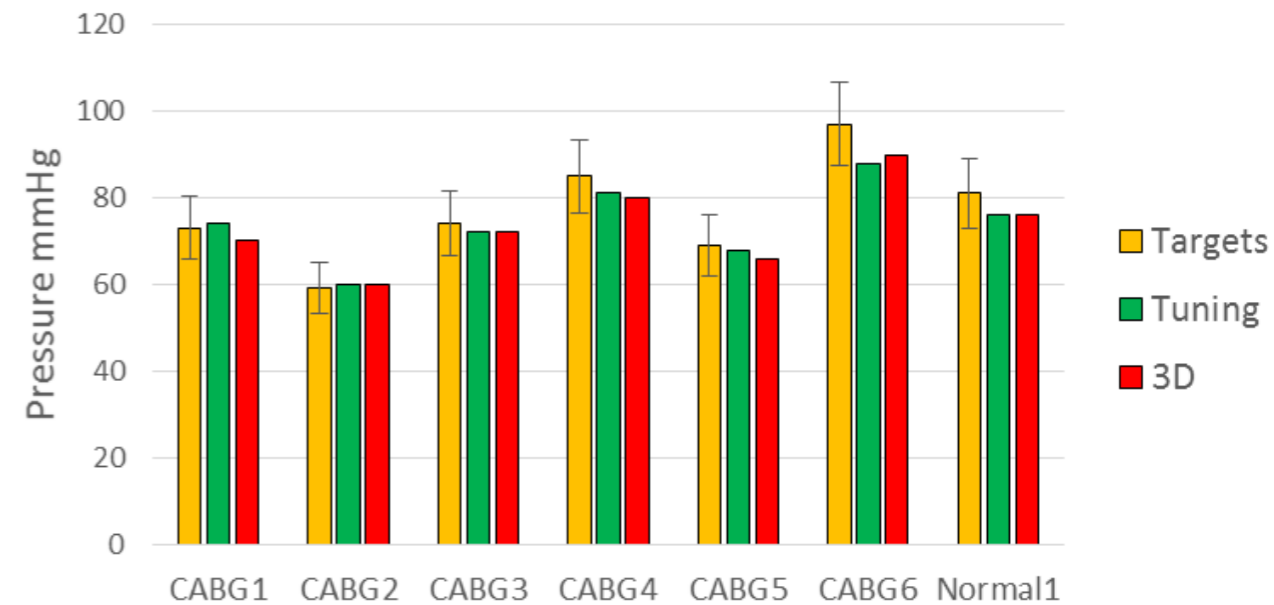
Automated Parameter Tuning



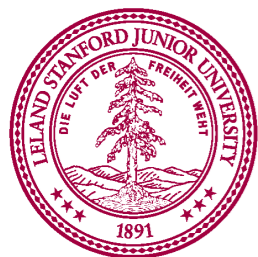
Stroke Volume



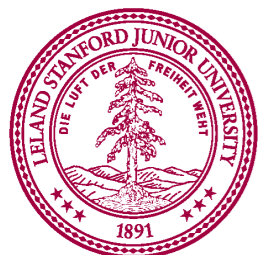
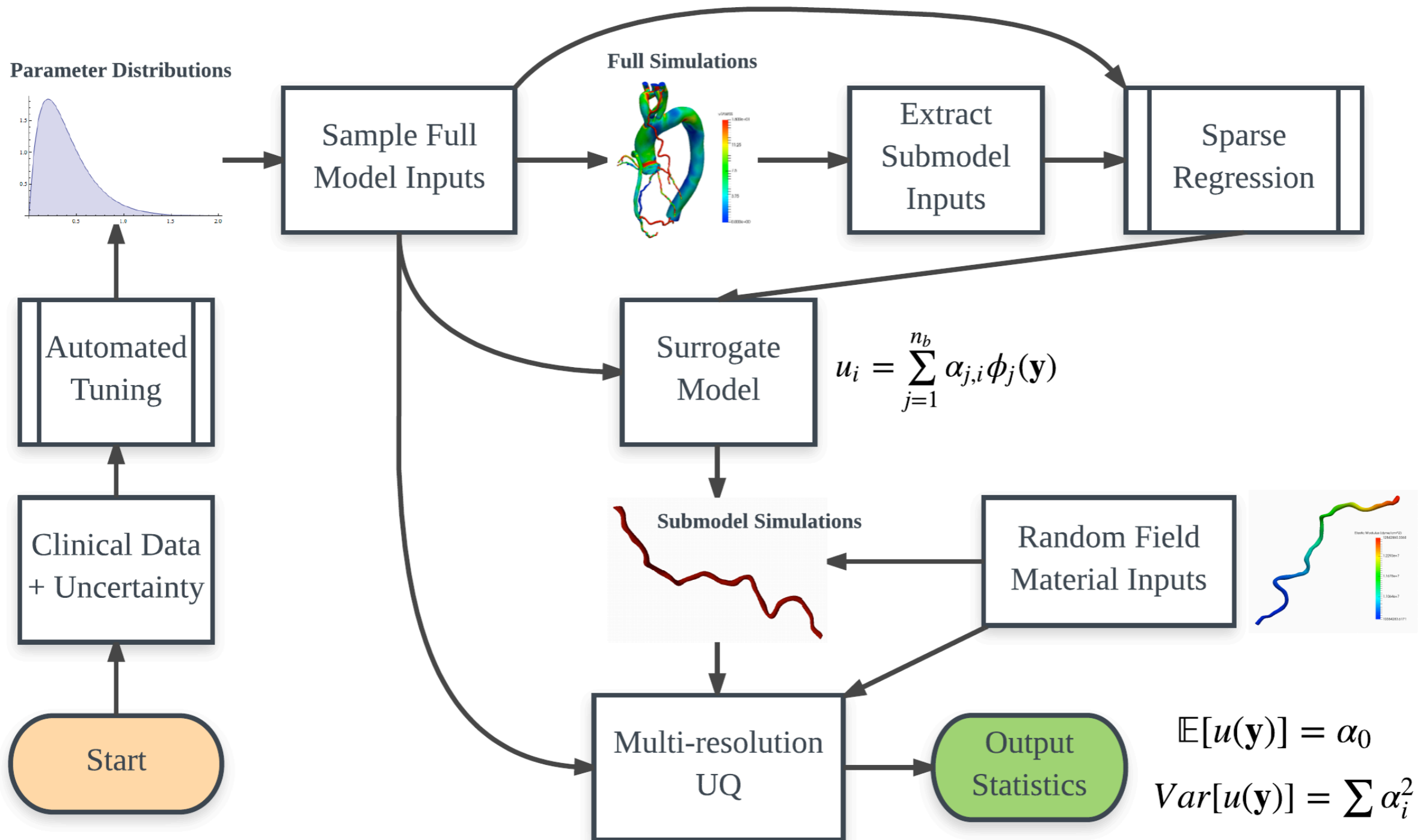
Pao Diastolic



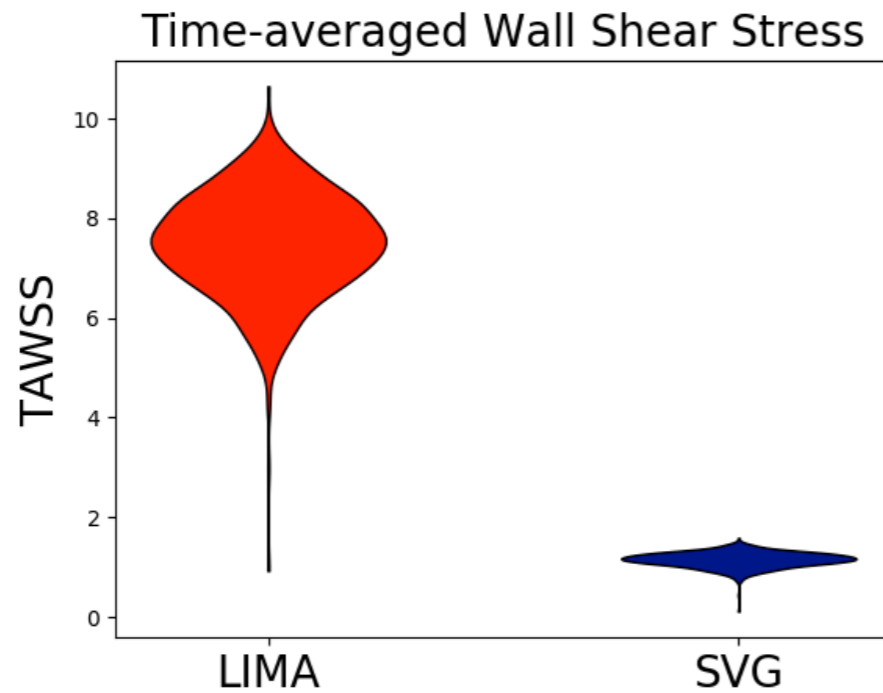
Tran, J. S., Schiavazzi, D. E., Ramachandra, A. B., Kahn, A. M., Marsden, A. L., "Automated Tuning for Parameter Identification in Multiscale Coronary Simulations," Vol. 142(5), pp. 128–138, *Computers and Fluids*, (2017)



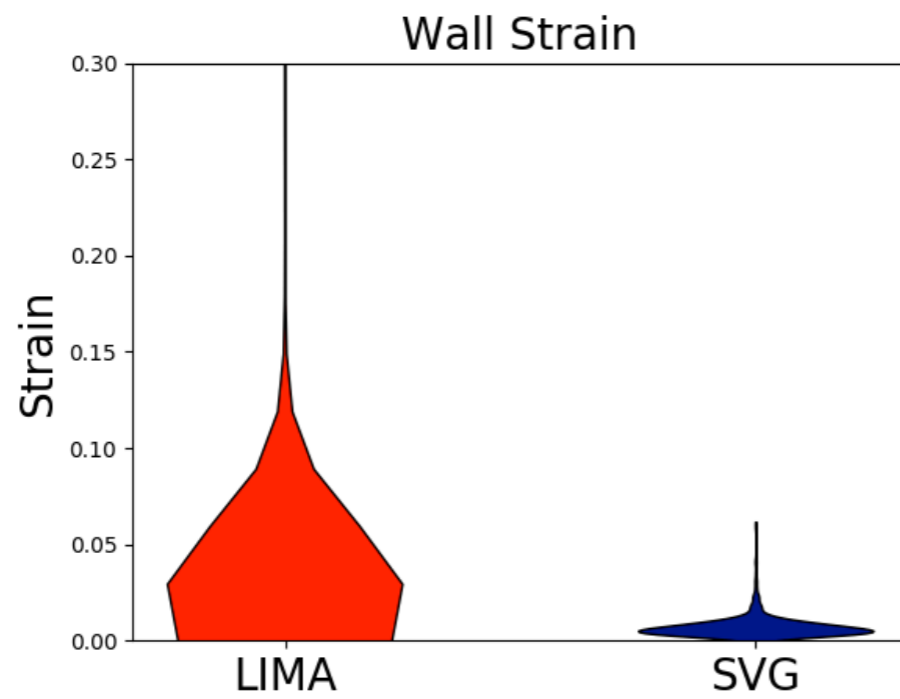
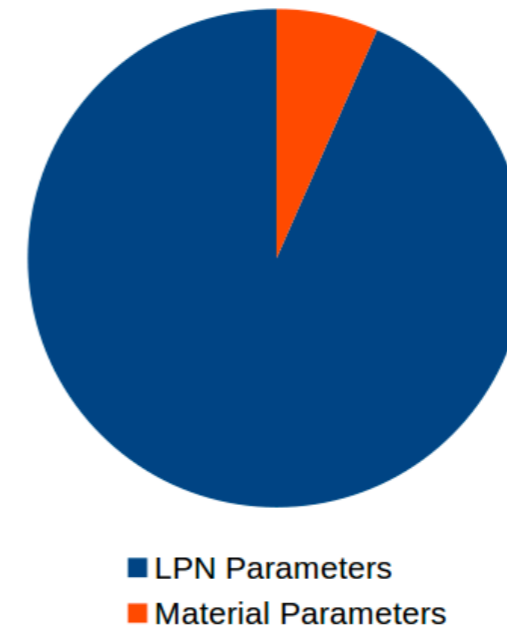
Uncertainty Propagation



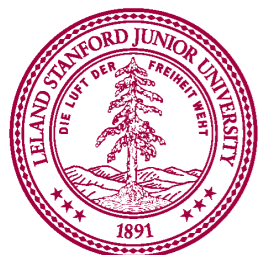
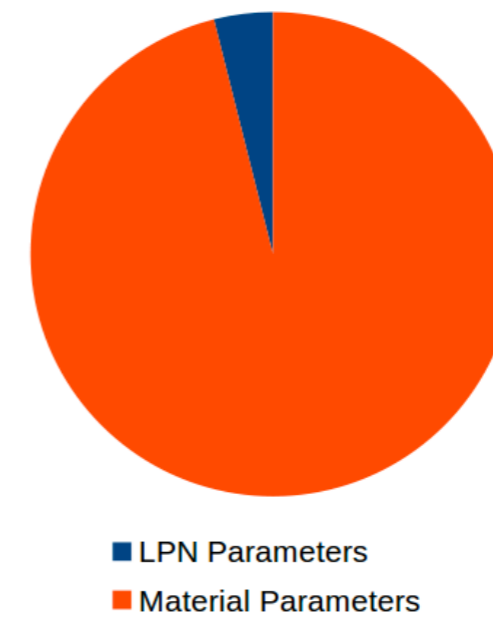
Statistics on model predictions



Variance Contributions to Time-average Wall Shear

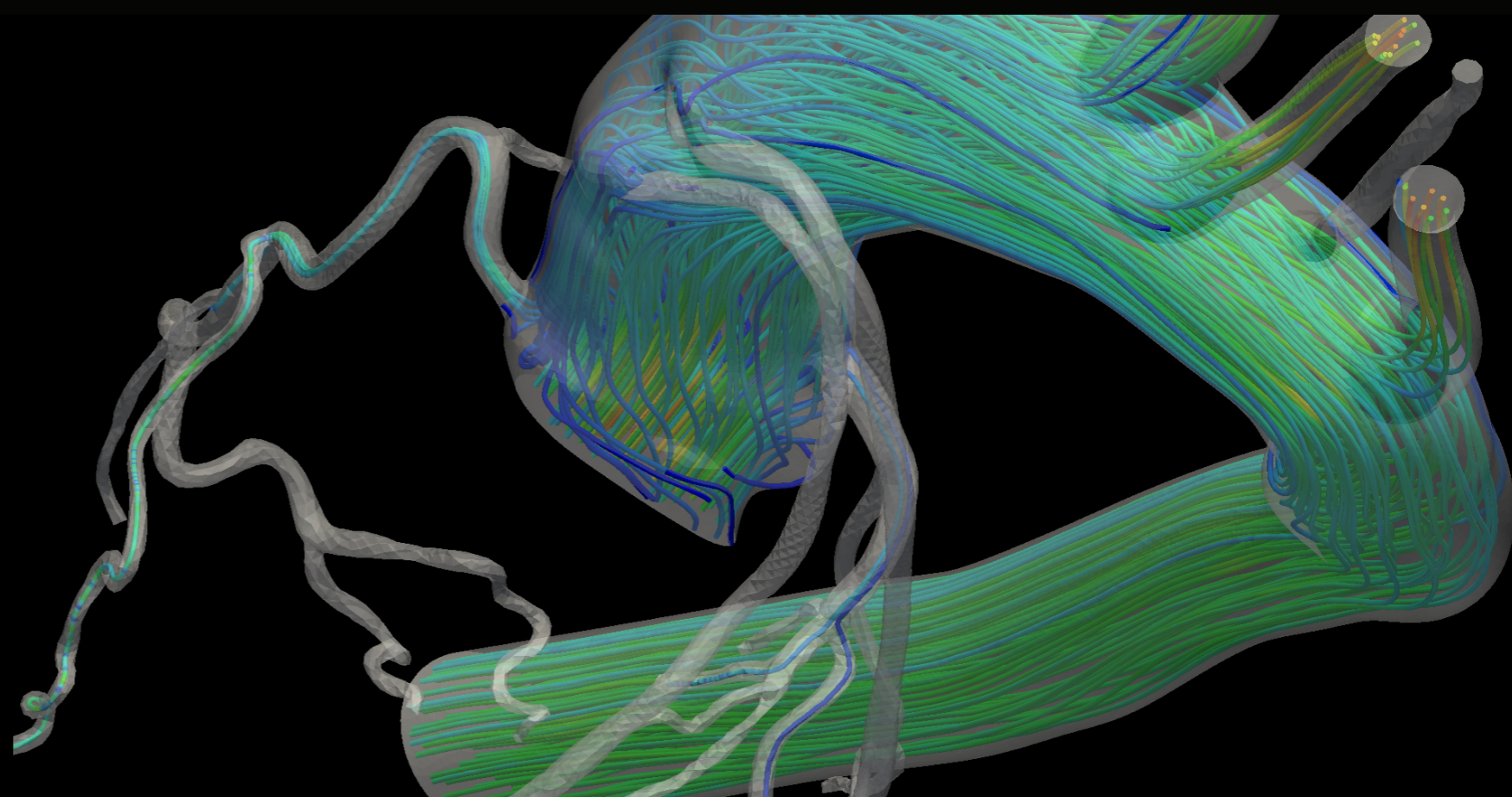


Variance Contributions to Wall Strain



Open Source

Providing tools to the research community



Alison Marsden
Stanford University

Shawn Shadden
UC Berkeley

Nathan Wilson
OSMSC

MAJOR FUNDING PROVIDED BY



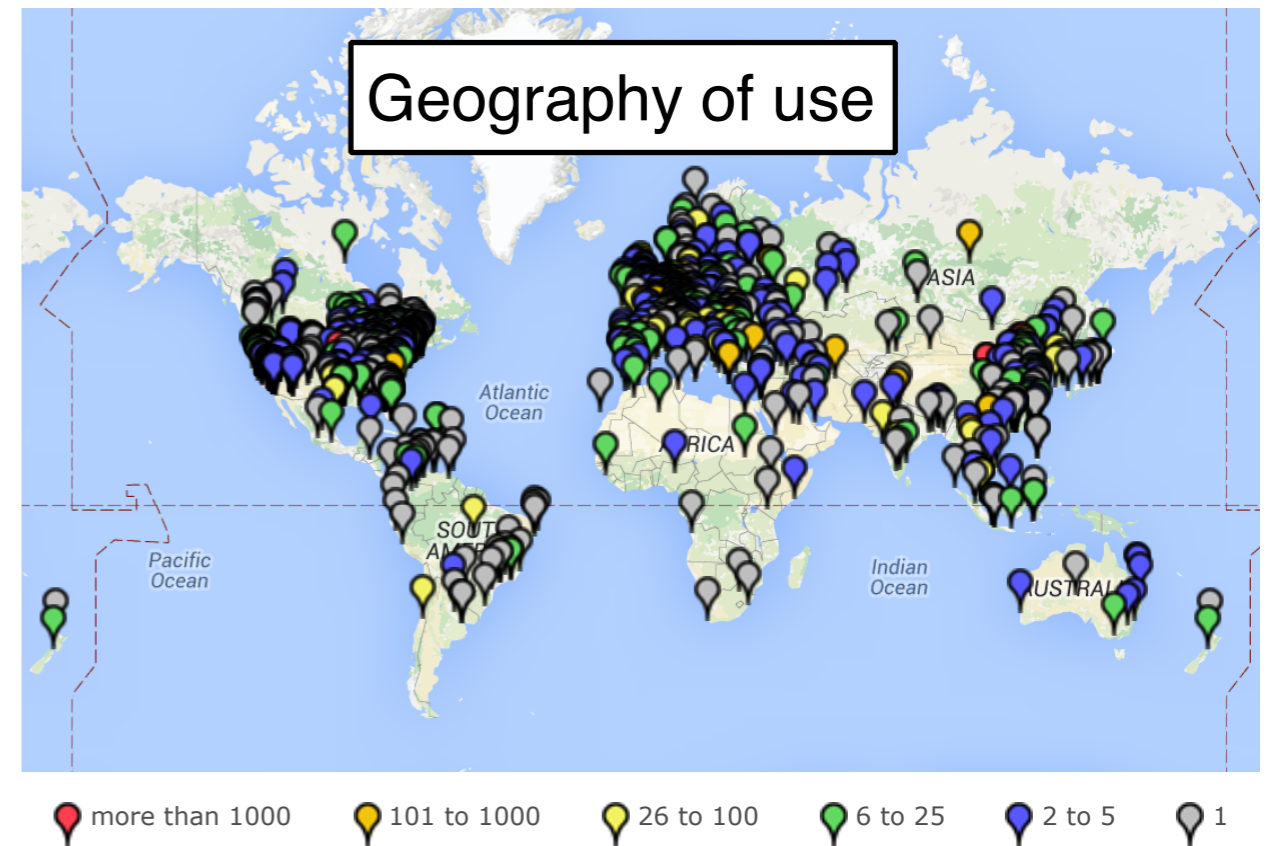
THE NATIONAL SCIENCE FOUNDATION

www.simvascular.org

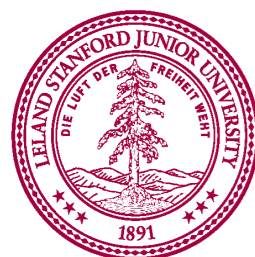
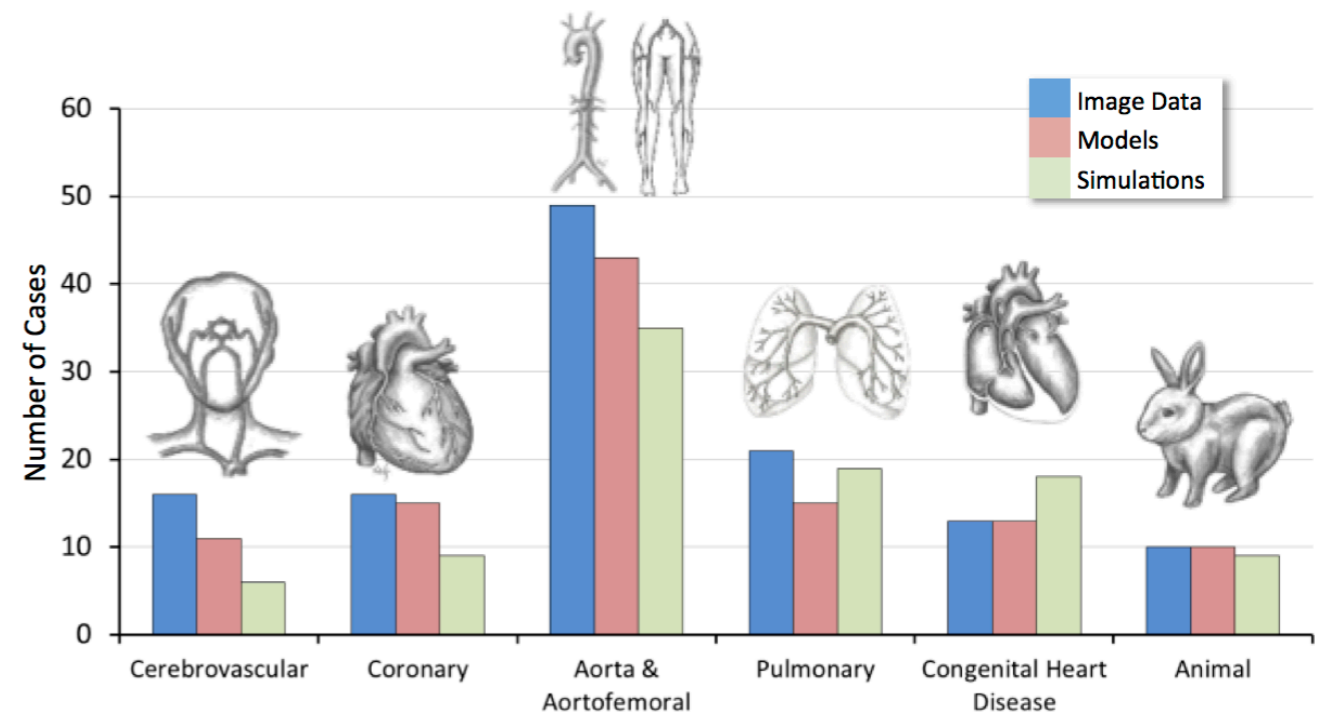
@SimVascular

Project Stats

- 1,959 unique users
- 4,866 unique downloads
- Google Scholar search for “SimVascular” produces ~140 publications/abstracts
- Used in coursework for project-based learning
- Vascular Model Repository provides 120 compatible data sets
- Source on GitHub
- Cross-platform support



vascularmodel.com



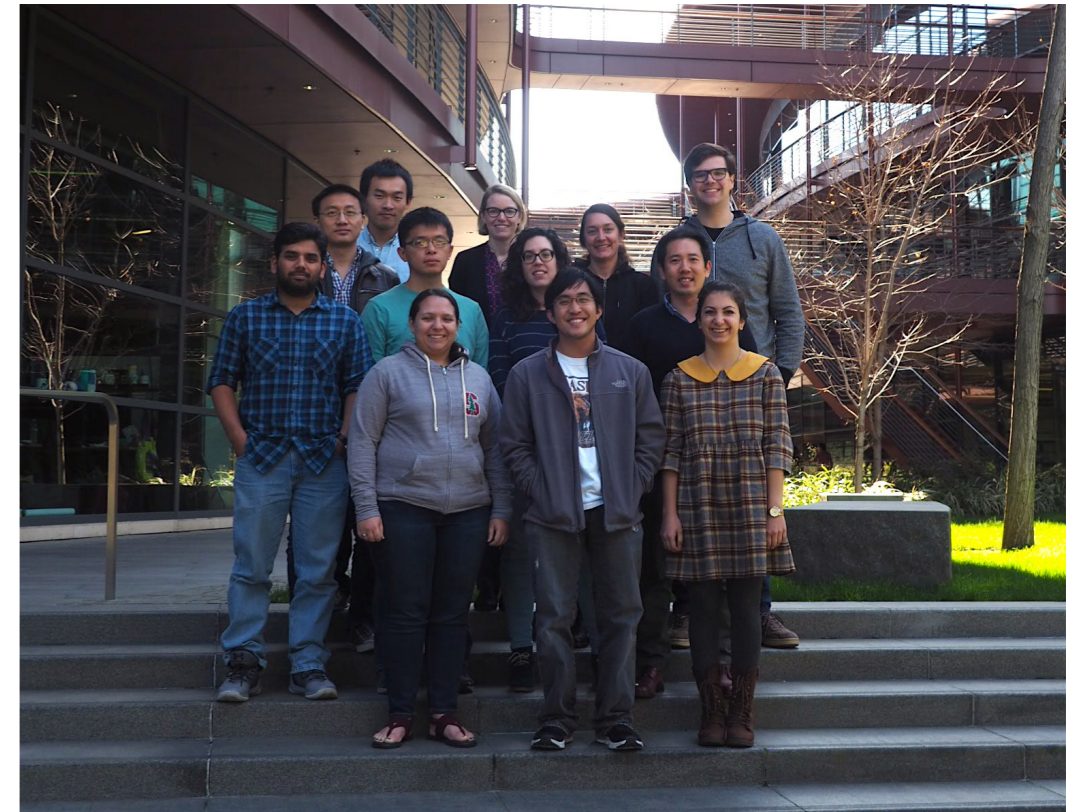
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Charles Audet, Ph.D.
Jay Humphrey, Ph.D.
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Richard Figliola, Ph.D.
Shawn Shadden, PhD.
Irene Vignon-Clementel, Ph.D.
Nathan Wilson, Ph.D.
Yuri Bazilevs, Ph.D.
Tzung Hsia, M.D.

Students/Postdocs:

Weiguang Yang, Ph.D.
Abhay Ramachandra
Daniele Schiavazzi, Ph.D.
Jameson Merkow
Gabriel Maher
Justin Tran
Noelia Grande Gutierrez
Aekaansh Verma
Hongzhi Lan, Ph.D.
Vijay Vedula, Ph.D.
Jessica Shang, Ph.D.
Mahdi Esmaily Moghadam, Ph.D.



Clinical Collaborators:

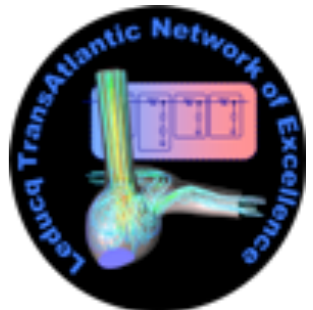
Jeffrey A. Feinstein, M.D.
Tain-Yen Hsia, M.D.
Jane C. Burns, M.D.
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Modeling Of Congenital Hearts Alliance (MOCHA)



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