



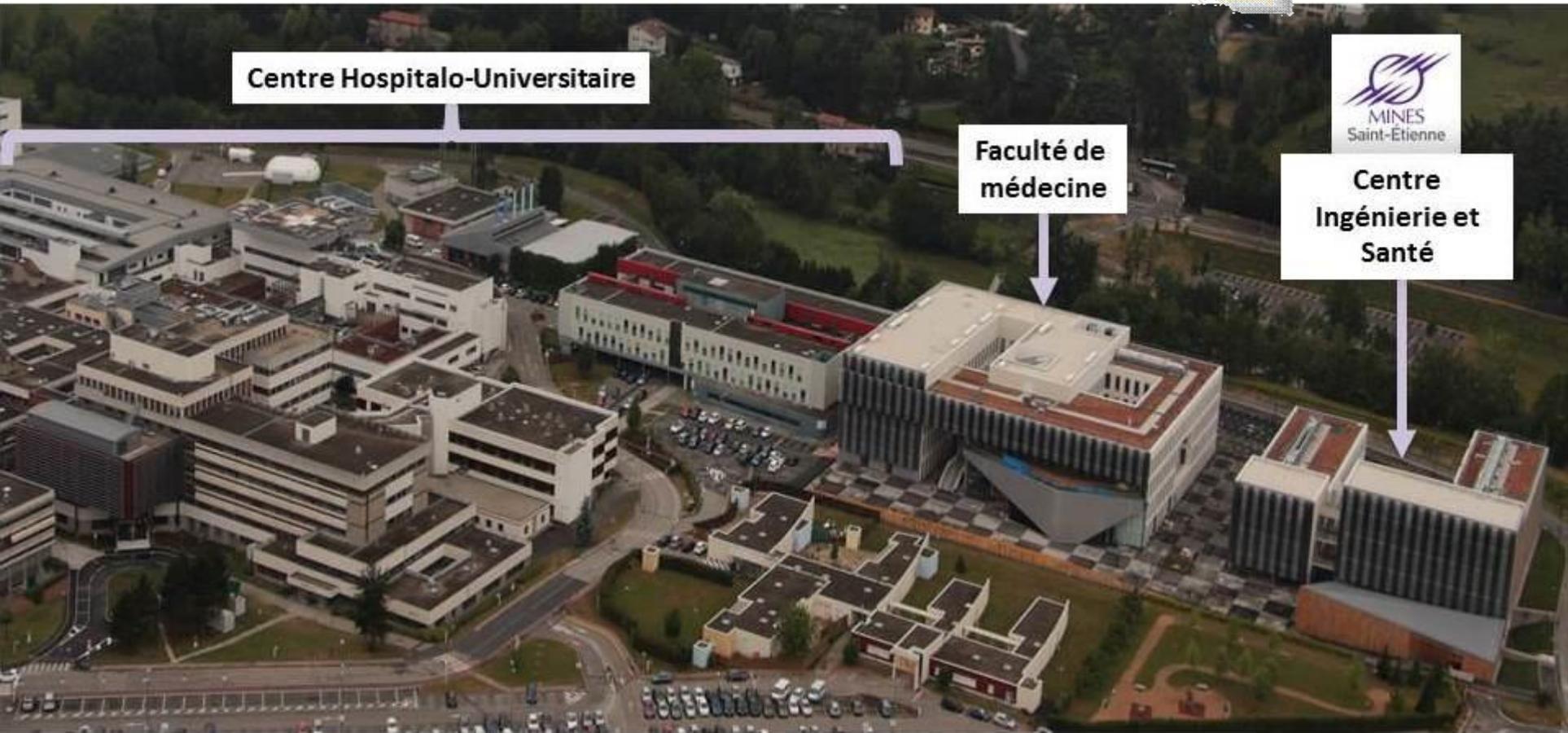
Continuum mechanics in cardiovascular engineering and biology



Prof. Stéphane AVRIL



MINES SAINT-ETIENNE
First Grande Ecole
outside Paris
Founded in 1816



Centre Hospitalo-Universitaire

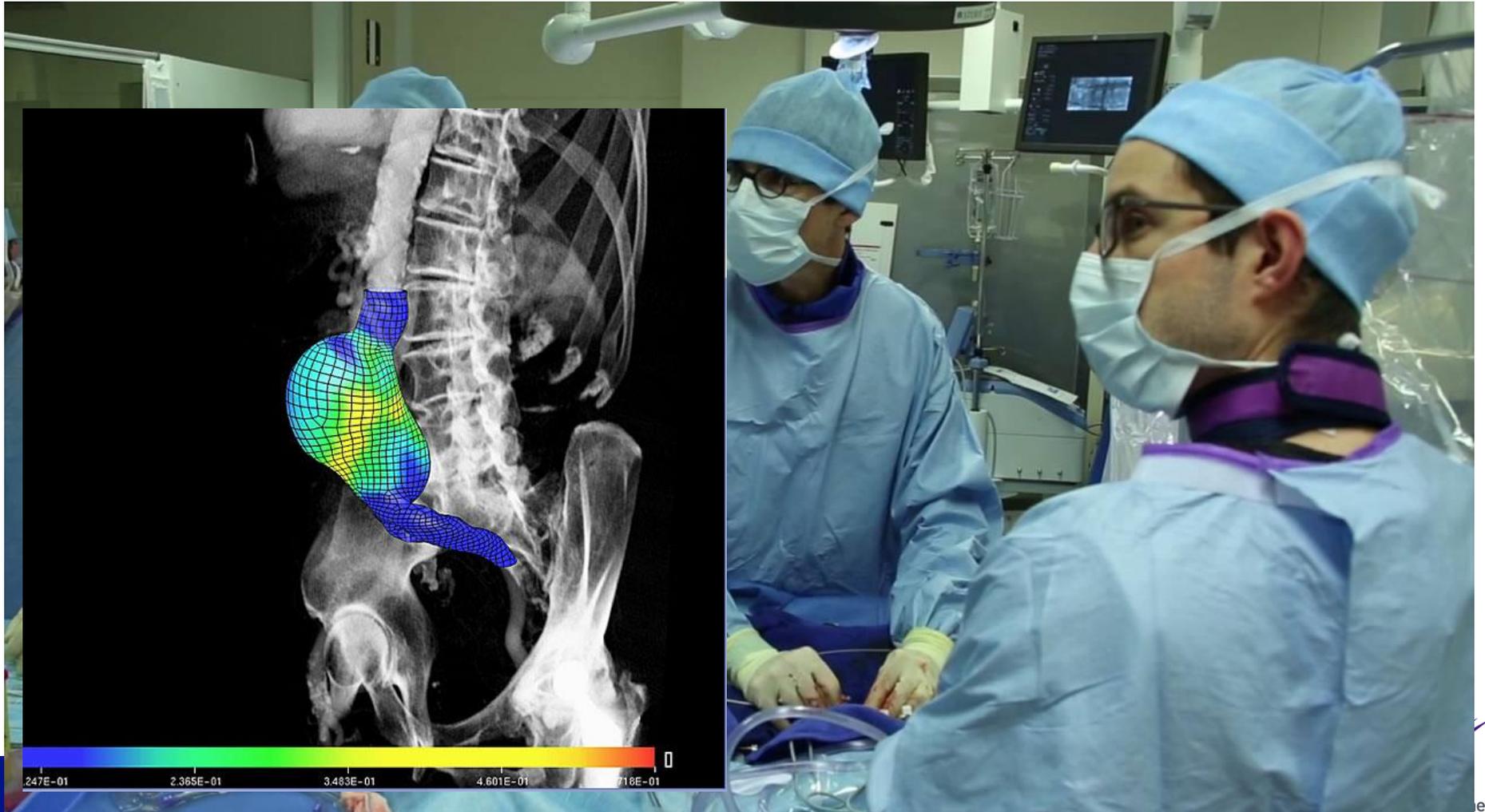
**Faculté de
médecine**

**Centre
Ingénierie et
Santé**



Computational mechanics in the OR for vascular surgery?

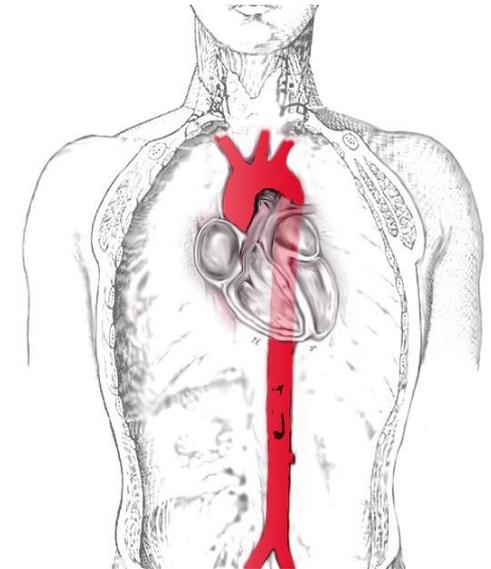
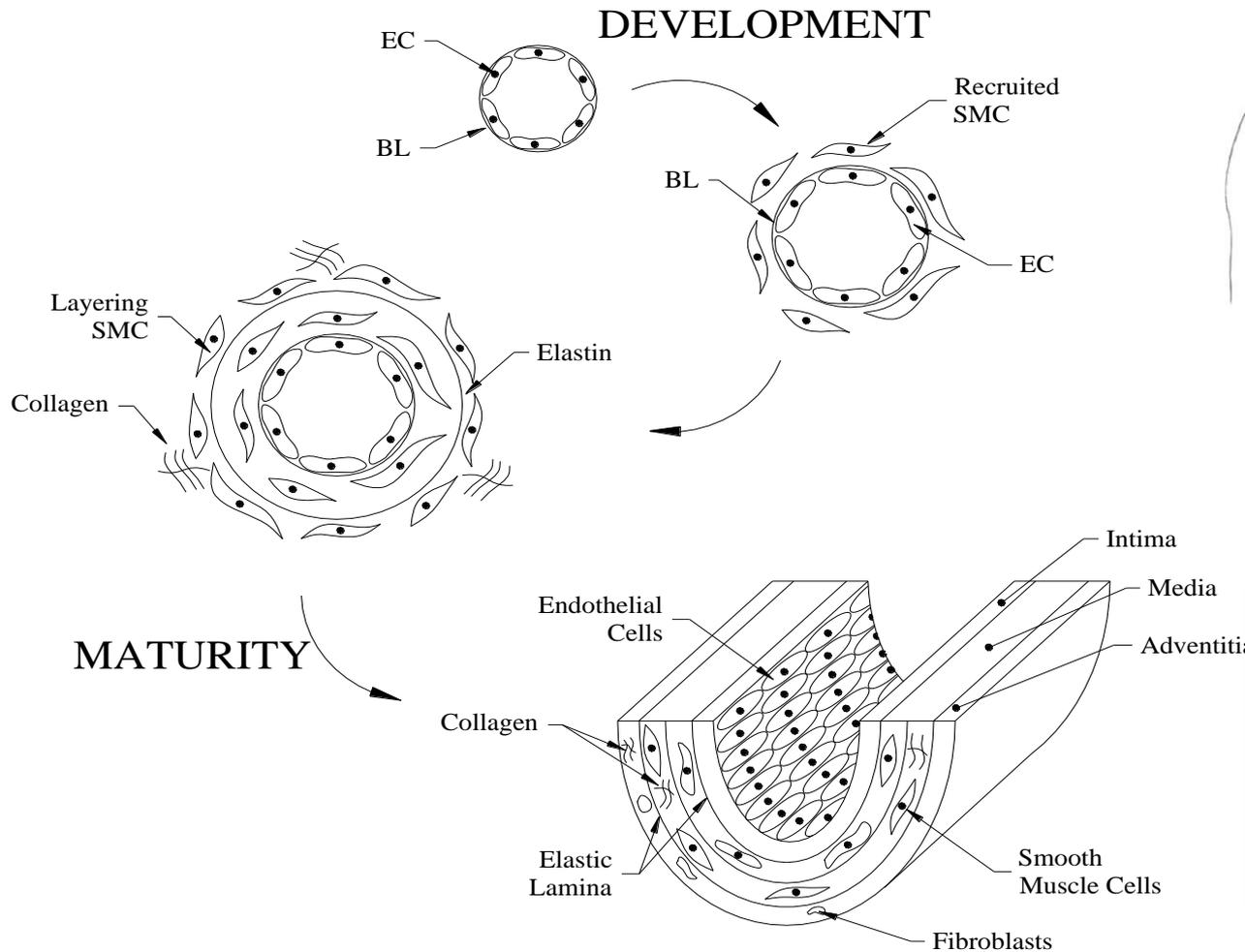
www.predisurge.com



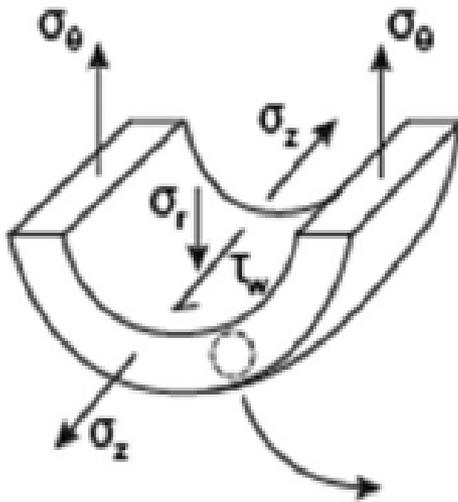


Arterial biomechanics and mechanobiology – Position of photomechanics

Schematic representation of aortic structure



Basics of aortic mechanics

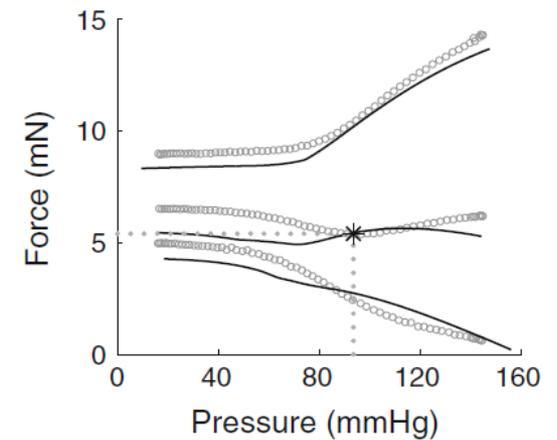
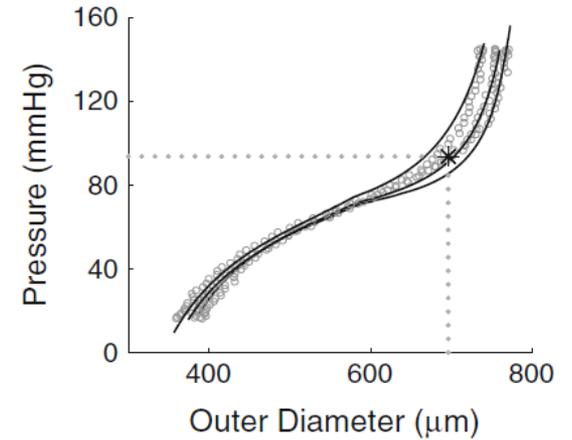
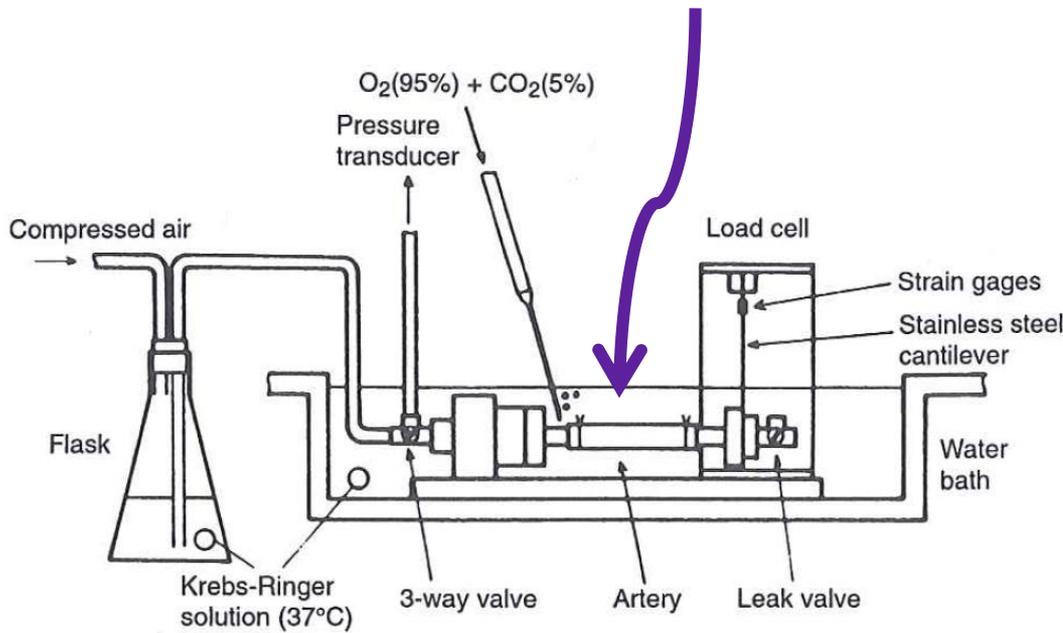


$$\tau_w = \frac{4\mu Q}{\pi a^3}, \quad \sigma_\theta = \frac{P a}{h}$$

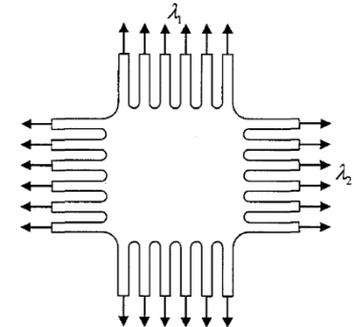
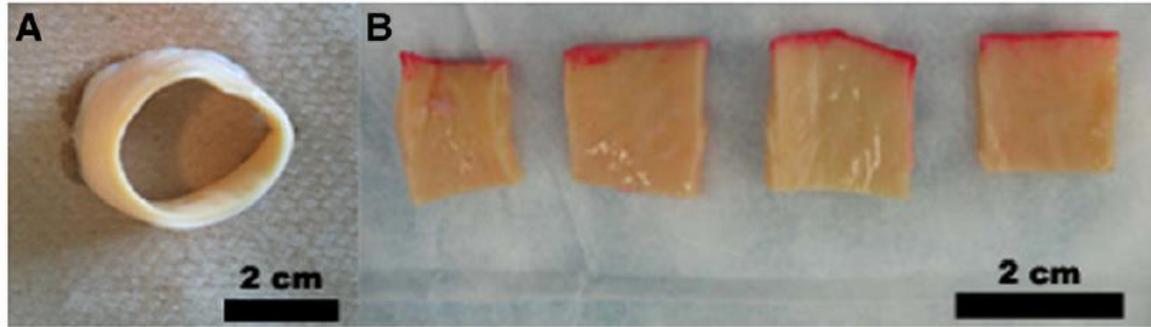
$$\sigma_z = \frac{f_z}{\pi (b^2 - a^2)} = \frac{f_z}{\pi h (2a - h)}$$

Humphrey JD (2002) *Cardiovascular Solid Mechanics: Cells, Tissues, and Organs*, Springer-Verlag, NY

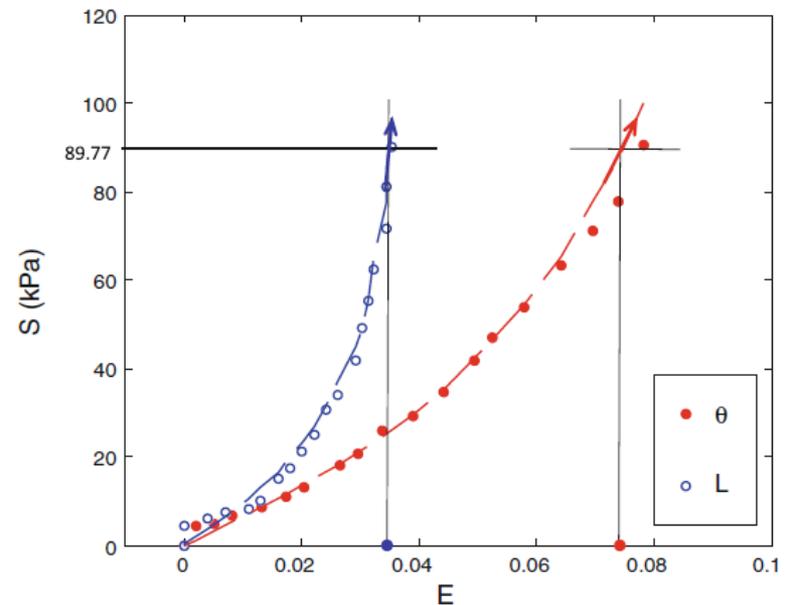
Functional biomechanical behavior



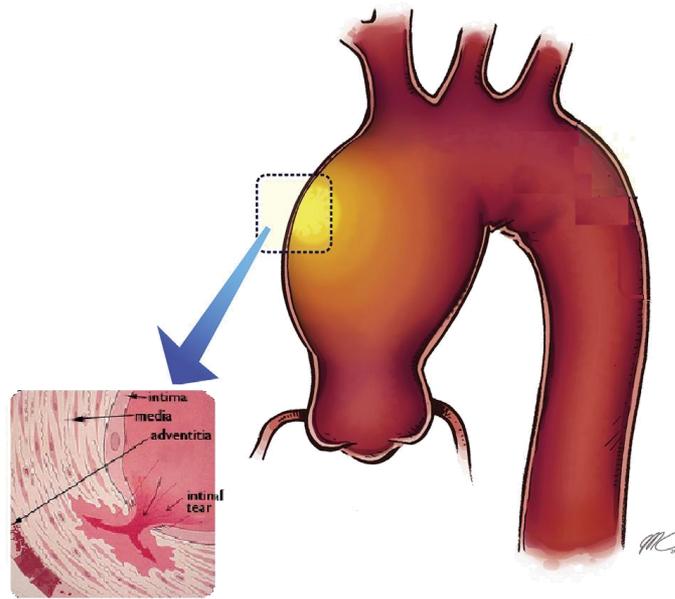
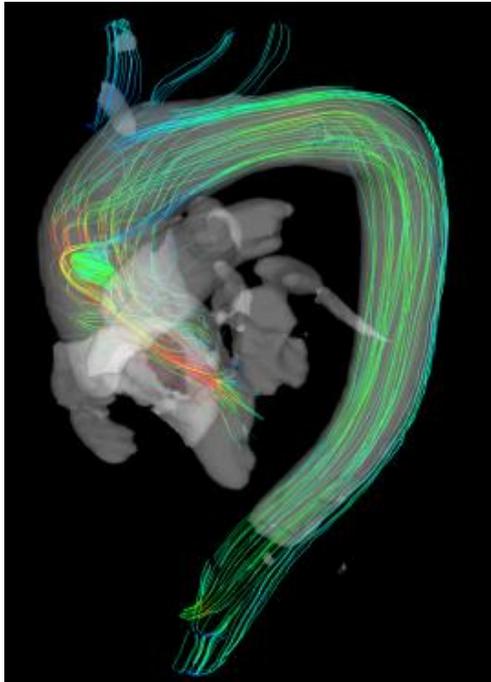
Material characterization and constitutive modeling



$$W = C_{10} (\bar{I}_1 - 3) + \frac{1}{D} \left(\frac{J^2 - 1}{2} - \ln J \right) + \frac{k_1}{2k_2} \sum_{\alpha=1}^N \left\{ \exp \left[k_2 \langle \bar{E}_\alpha \rangle^2 \right] - 1 \right\}$$



Prediction of risk of rupture and dissection



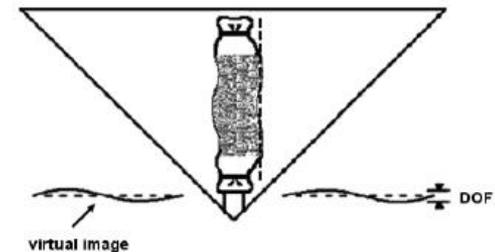
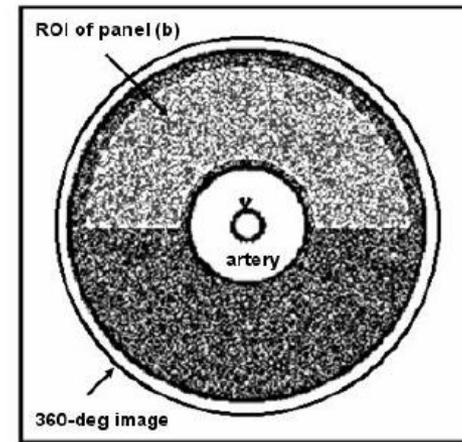
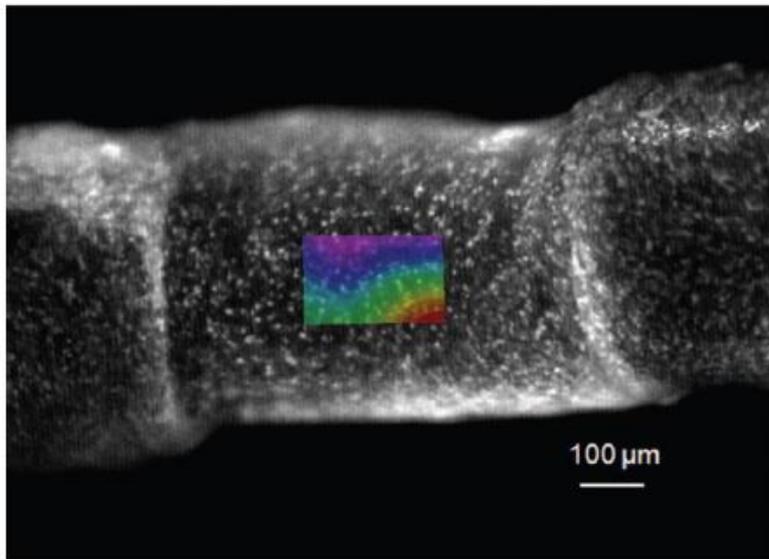
MEASUREMENT OF THE RESPONSE USING DIGITAL IMAGE CORRELATION



classical



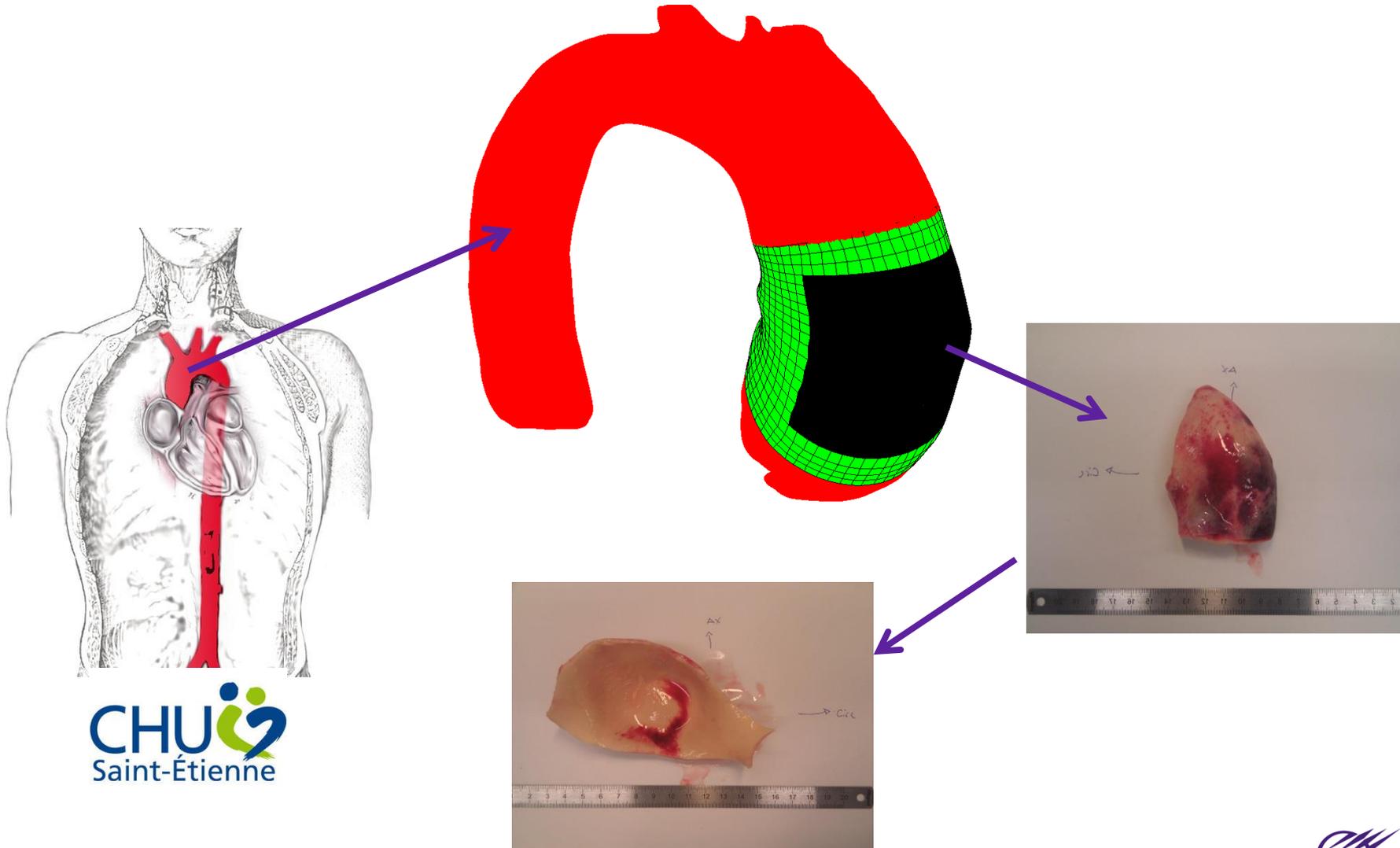
panoramic



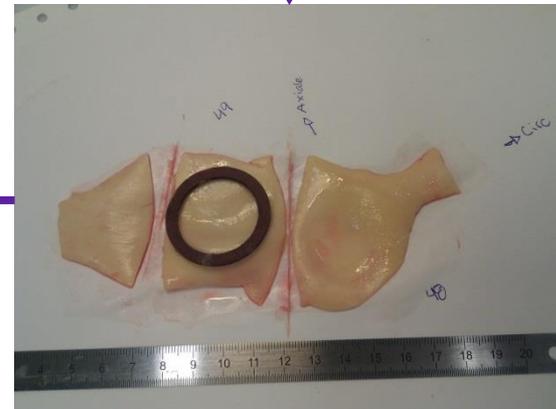
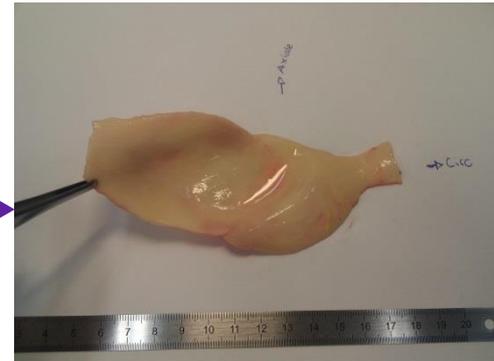
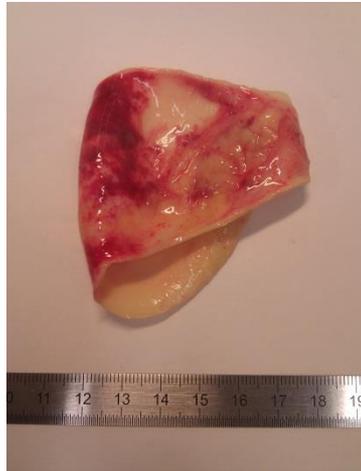
Badel et al. CMBBE, 15, p 37-48, 2012.

Genovese. Optics Lasers Eng, 47, p 995-1008, 2009.

Collection of the samples

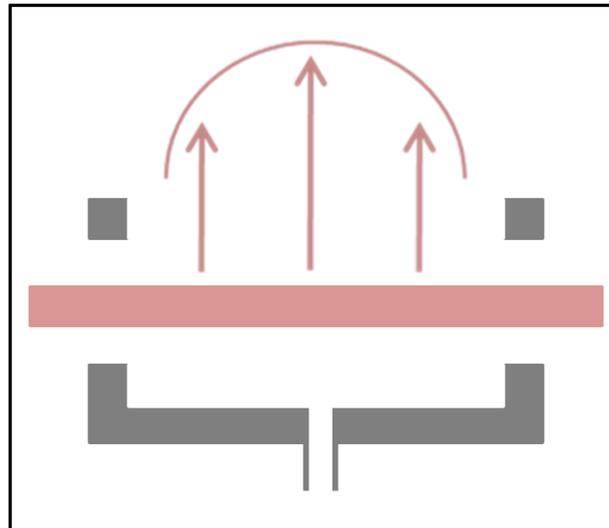


Preparation

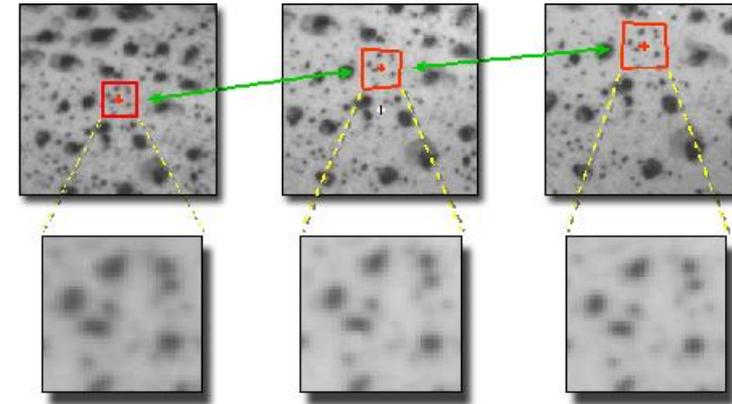


Bulge inflation test

Romo et al. Journal of Biomechanics -2014.

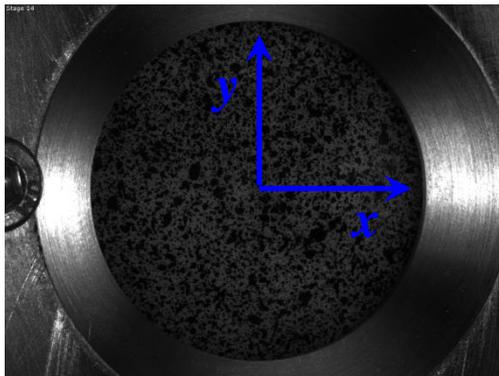


Full-field measurements using sDIC

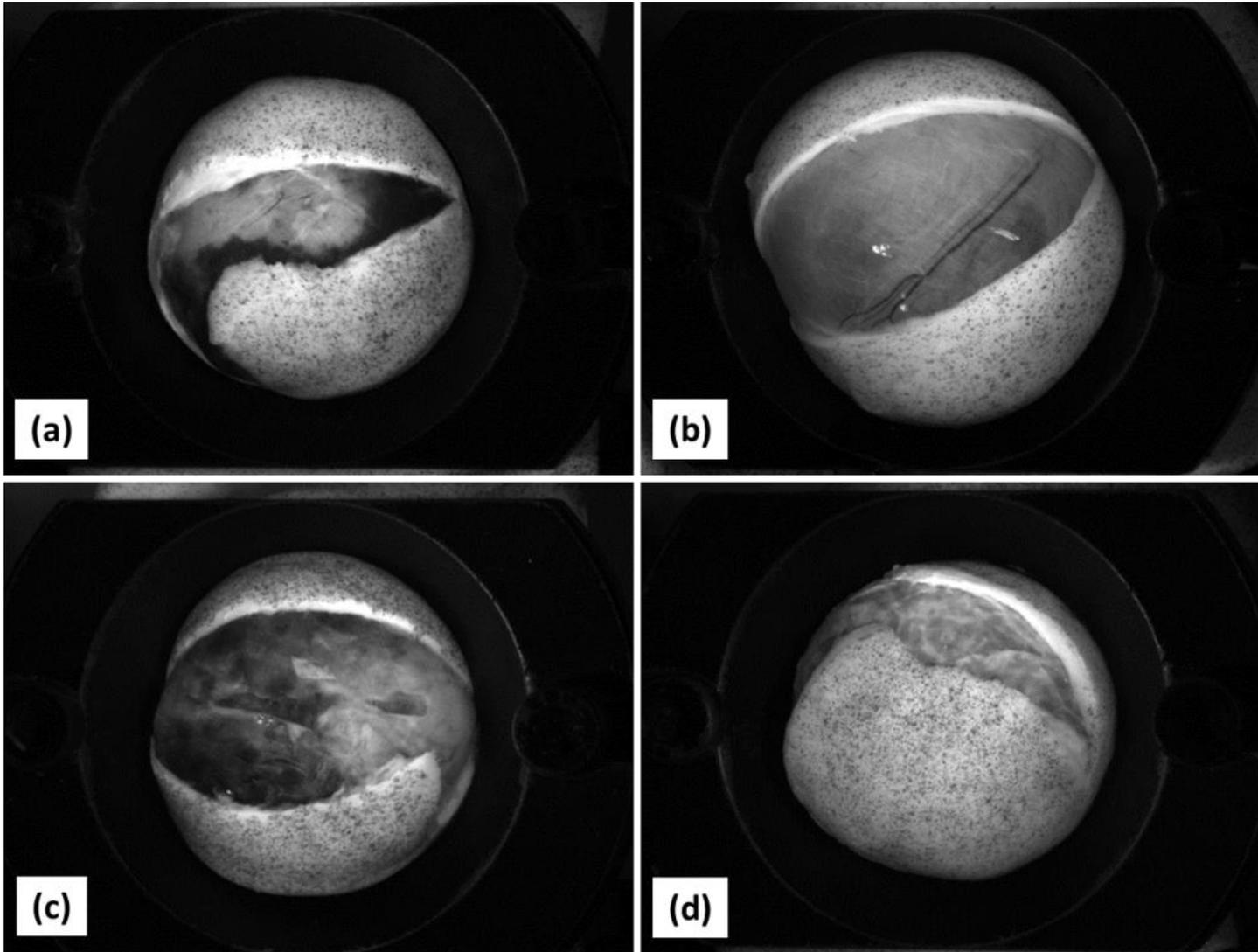


Undeformed

Deformed



Rupture profiles



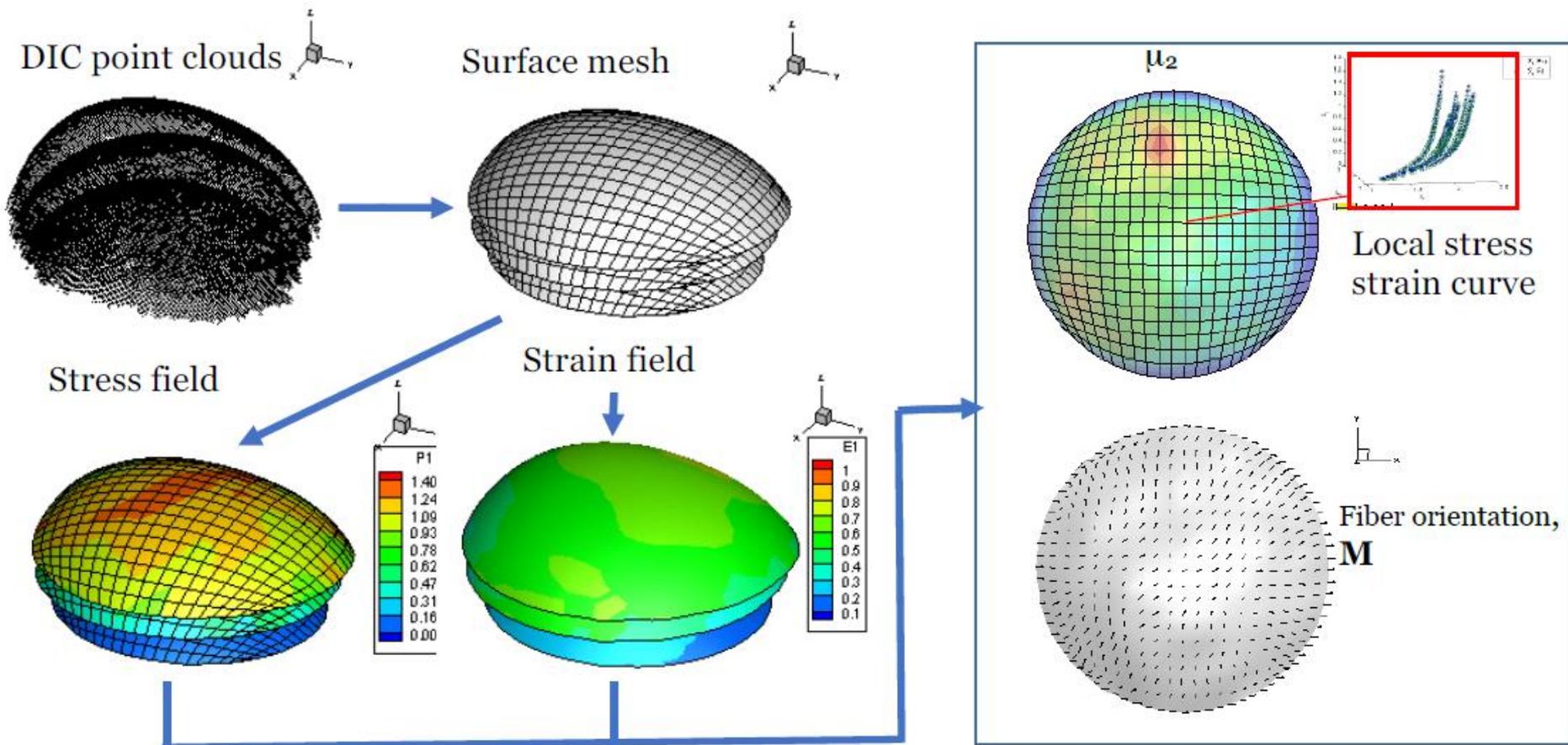
Blood flow

50% of aortas ruptured with an angle θ equal to 90°

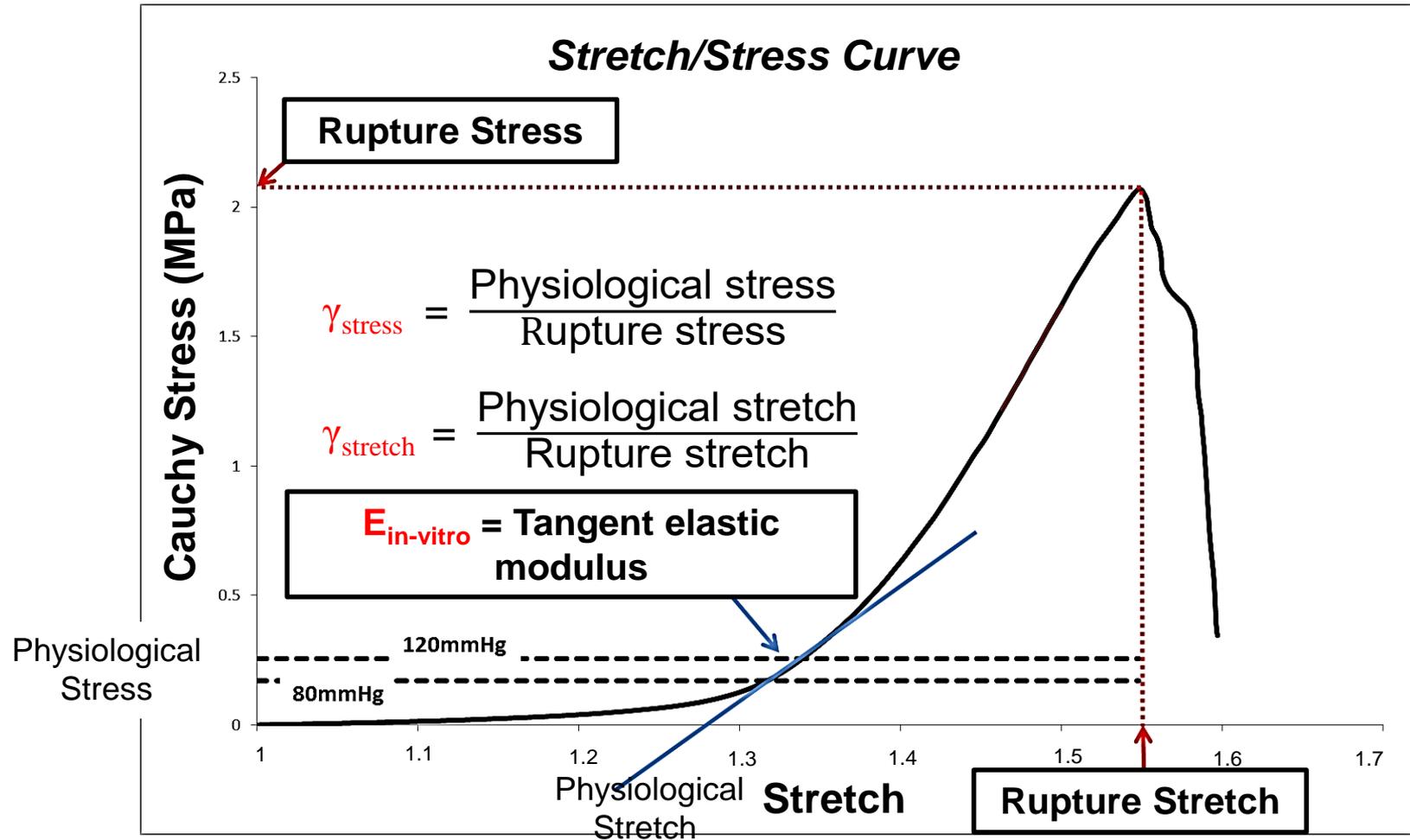


Identification of local material properties

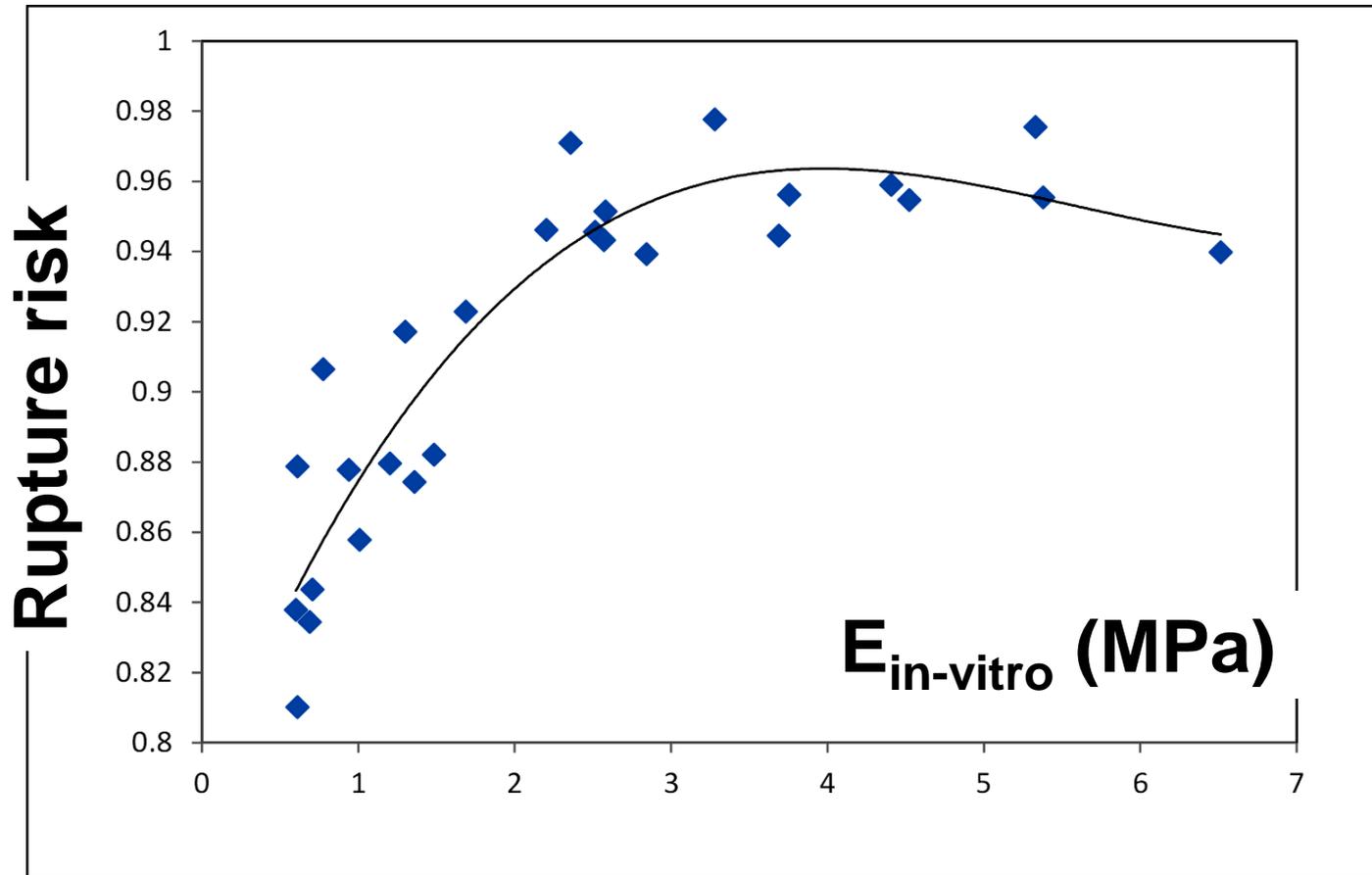
Davis et al. BMMB – 2015.
Davis et al. JMBS – 2016
Zhao et al. Acta Biomaterialia - 2016



Rupture risk estimation



Correlation between the stretch-based rupture risk and the tangent elastic modulus



Duprey A, et al. Biaxial rupture properties of ascending thoracic aortic aneurysms. *Acta Biomaterialia* 2016.



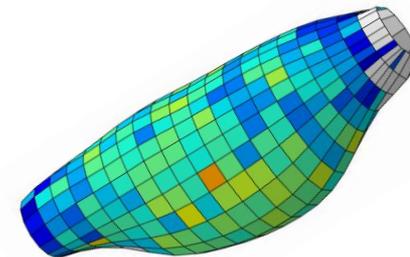
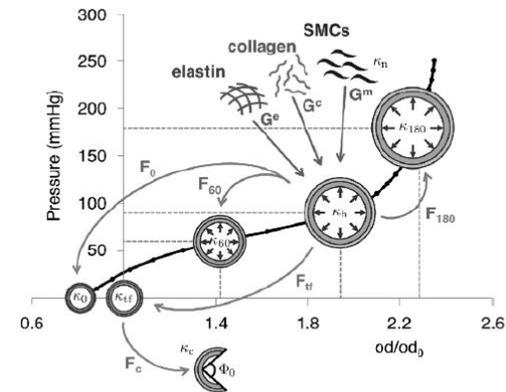
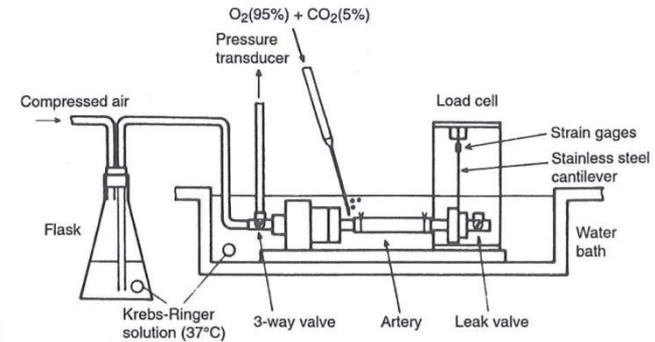
Understanding aneurysm growth using mechanobiology and photomechanics

Altered mechanics induce biological responses, including gene expression, protein activation and cell phenotype



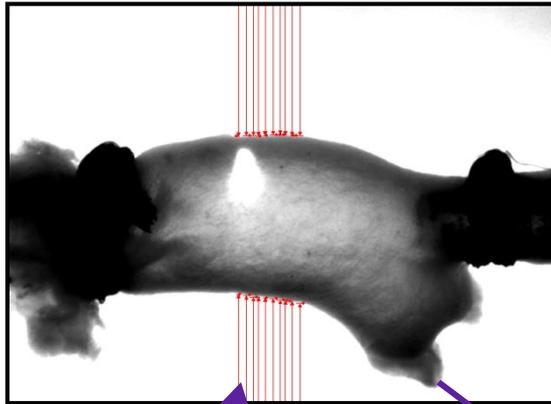
APPROACH

1. Experiments
2. Material model
3. Inverse method

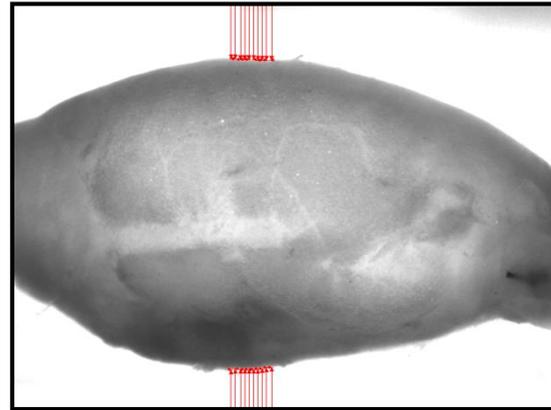


Study Design

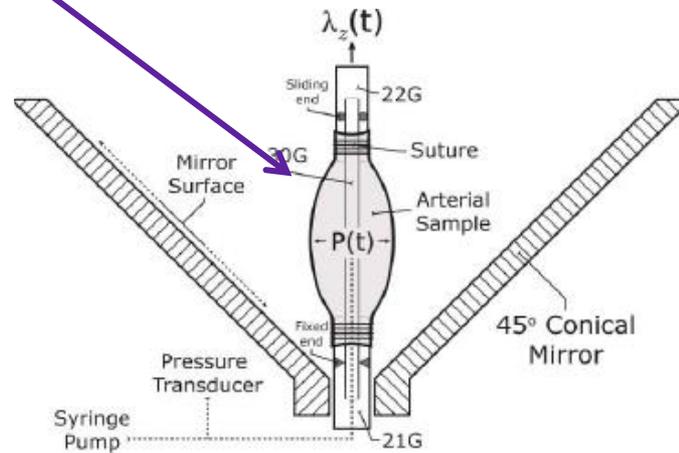
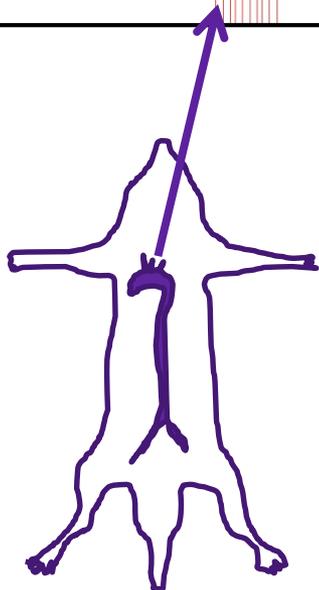
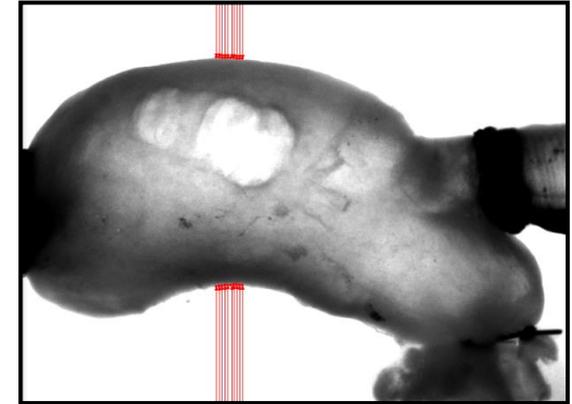
Control



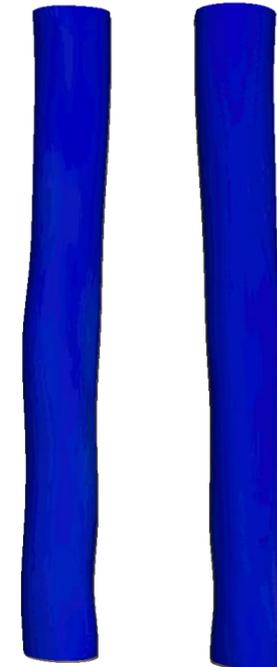
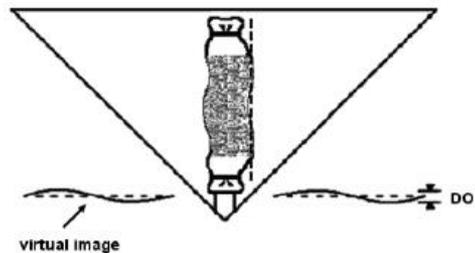
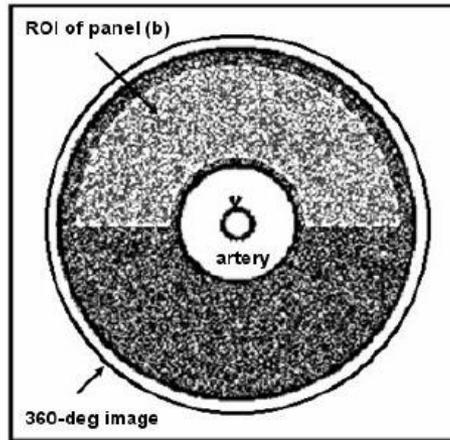
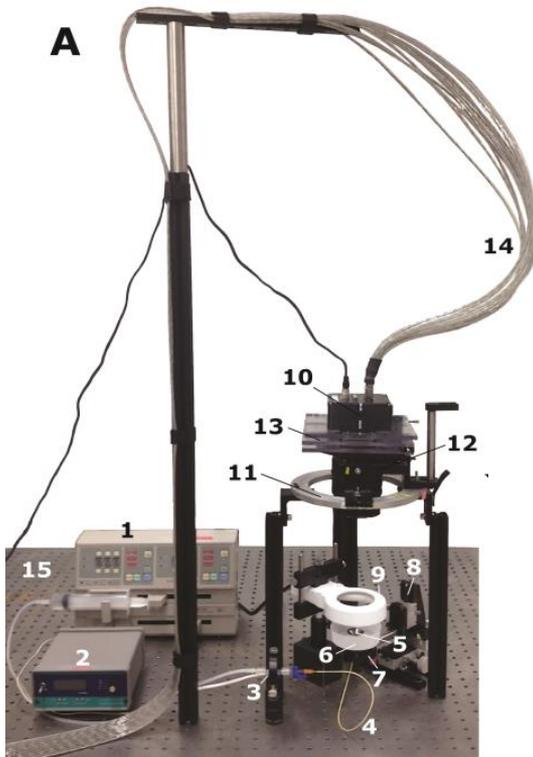
Fibulin 4 SMC KO



Fibrillin 1 *mgR/mgR*



The pDIC technique



Posterior

Anterior

pDIC measurements

Fibulin 4 SMC KO

Fibrillin 1 *mgR/mgR*

ventral

dorsal

inflation

ventral

dorsal

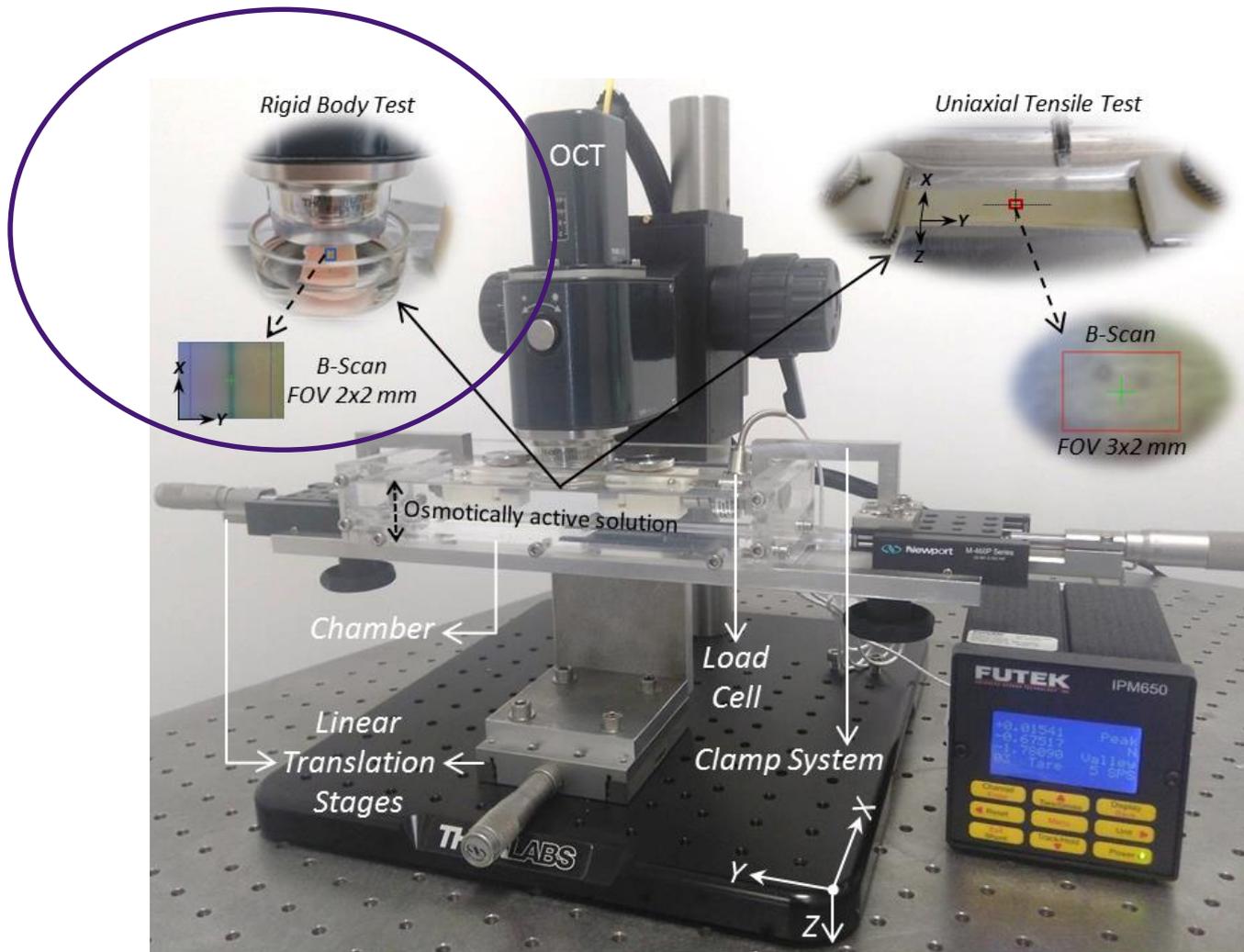
inflation

6852

CS38



OCT-DVC applied to arterial mechanics



Measurement of bulk deformation fields by Digital Volume Correlation on OCT images

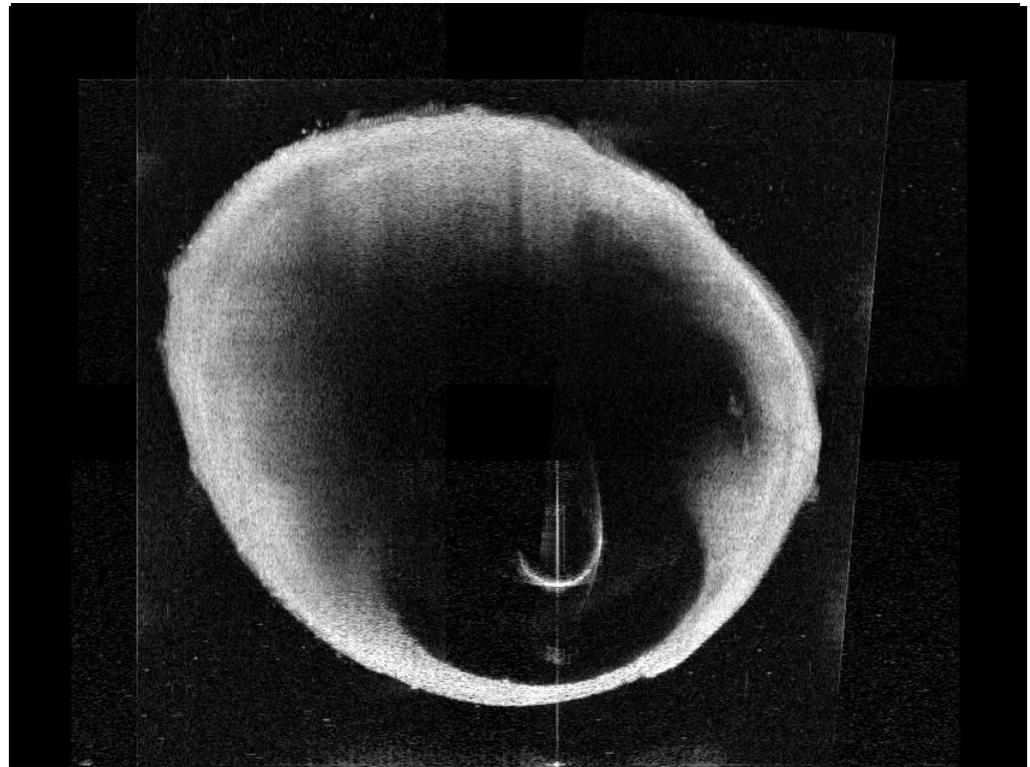
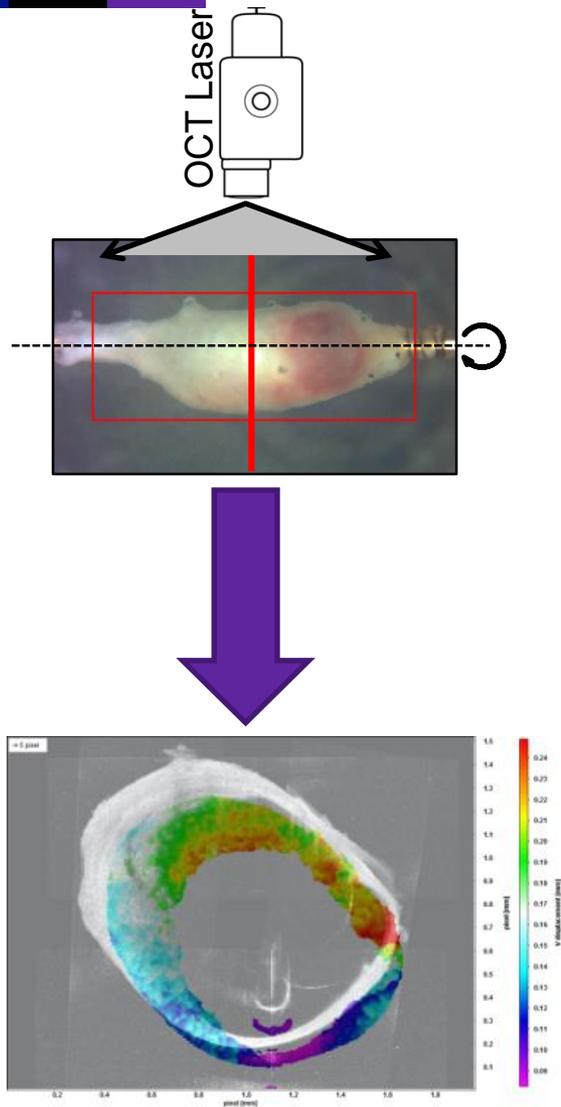
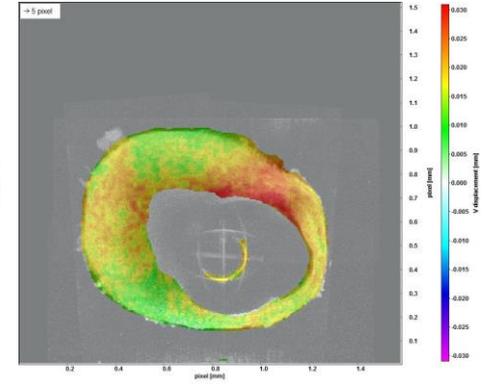
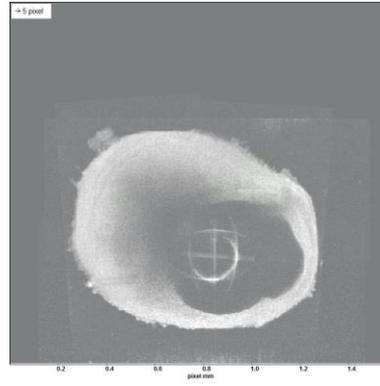
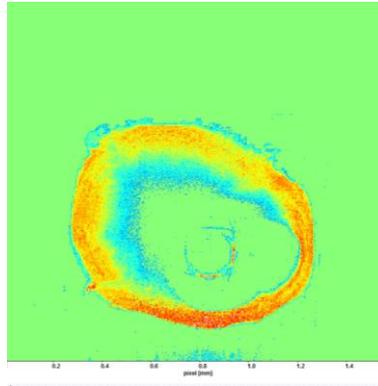
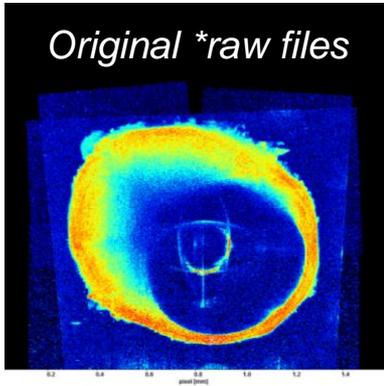


Image Processing Methodology – Inflation Test



Title: REF

Width: 667 pixels

Height: 640 pixels

Depth: 100 pixels

Voxel size: 1x1x1 pixel³ (2.38um)

Original
Image Parameters

Pressures: 20, 40, 60, 80, 100, 120, 140

Lambda (λ) 1 – 2 – 3

- **Algorithmic Mask – Threshold: 48**
- **Correlation window size [voxel]: 24 (8x8x8)**
- **Overlap : 75%**
- **Passes : 10**
- **Required valid voxel per window: 44%**

Correlation window sizes (setup up to six steps):

	Size [voxel]	Shape	Overlap [%]	Peak search radius [voxel]	Volume Binning	Passes
Step 1	24	1:1	0	8	8x8x8	2
Step 2	20	1:1	0	4	4x4x4	2
Step 3	16	1:1	0	2	2x2x2	2
Step 4	12	1:1	0	1	no	2
Step 5	10	1:1	75	1	no	2
Step 6	8	1:1	75	1	no	10

Intensity threshold for compression: 0 counts
(only GPU: 0 counts <=> lossless)

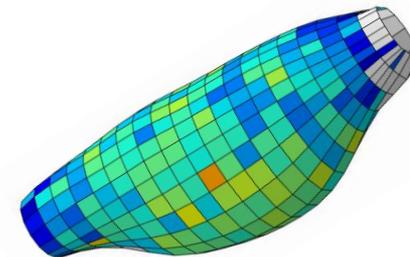
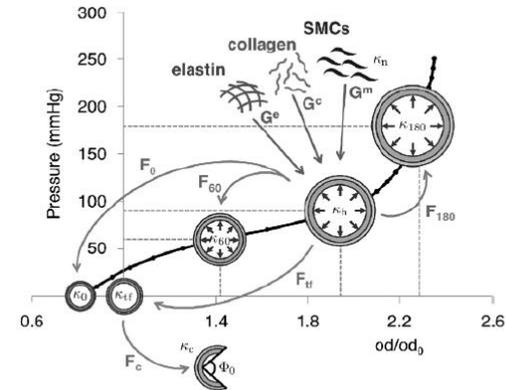
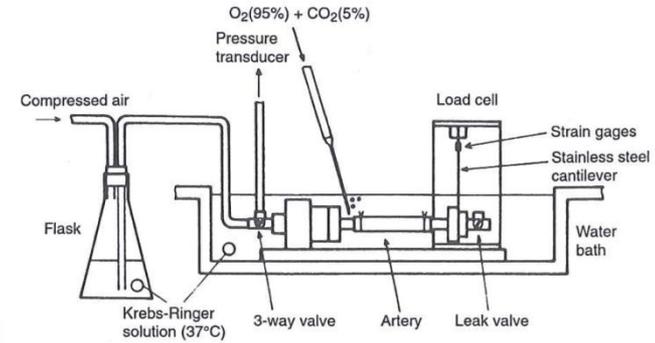
Required valid voxel per window: 45 %

Algorithmic mask with operation pipeline:

<input type="checkbox"/> local Stdev	over N pixel, N=	3
<input type="checkbox"/> sliding maximum	filter length N pixel, N=	4
<input checked="" type="checkbox"/> below threshold	set to 0, enter lower limit	48
<input type="checkbox"/> erosion	erode mask N times, N=	95
<input type="checkbox"/> above threshold	set to 0, enter upper limit	3
<input type="checkbox"/> erosion	erode mask N times, N=	105
<input type="checkbox"/> above threshold	set to 0, enter upper limit	1
<input type="checkbox"/> eliminate 0 pixel	set to !=1 (recommended !)	

APPROACH

1. Experiments
2. Material model
3. Inverse method



CONSTITUTIVE MODEL

Strain energy functions:

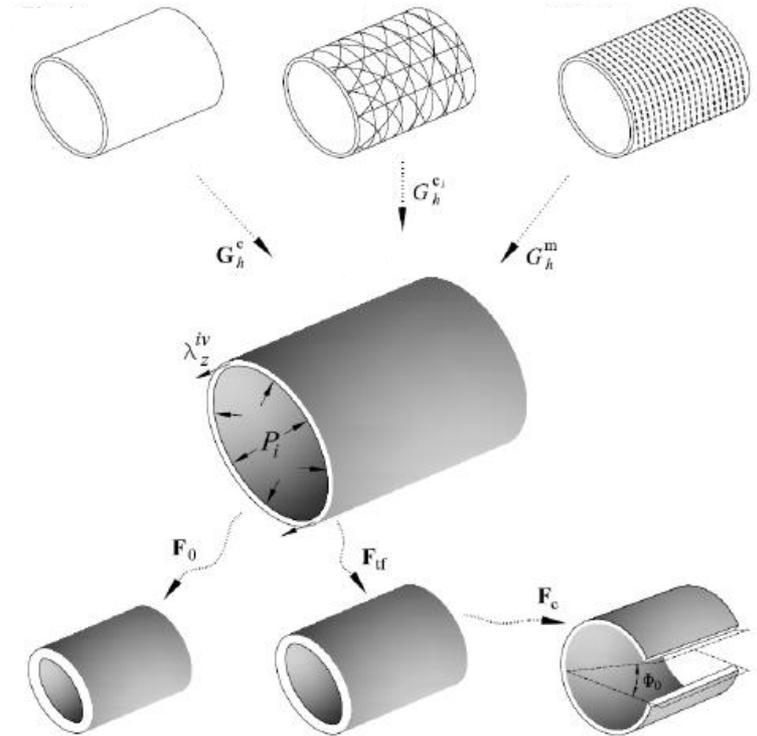
$$W = \phi^e W^e(\mathbf{F}^e) + \phi^m W^m(\lambda^m) + \sum_{j=1}^4 \phi^{c_j} W^{c_j}(\lambda^{c_j})$$

$$W^e(\mathbf{F}^e) = \frac{c^e}{2} \left[\text{tr} \left((\mathbf{F}^e)^T \mathbf{F}^e \right) - 3 \right]$$

$$W^m(\lambda^m) = \frac{c_2^m}{4c_3^m} \left[e^{c_3^m ((\lambda^m)^2 - 1)^2} - 1 \right]$$

$$W^c(\lambda^{c_j}) = \frac{c_2^c}{4c_3^c} \left[e^{c_3^c ((\lambda^{c_j})^2 - 1)^2} - 1 \right]$$

Bellini, et al., Ann. Biomed. Eng.,
42(3), pp. 488–502, 2014



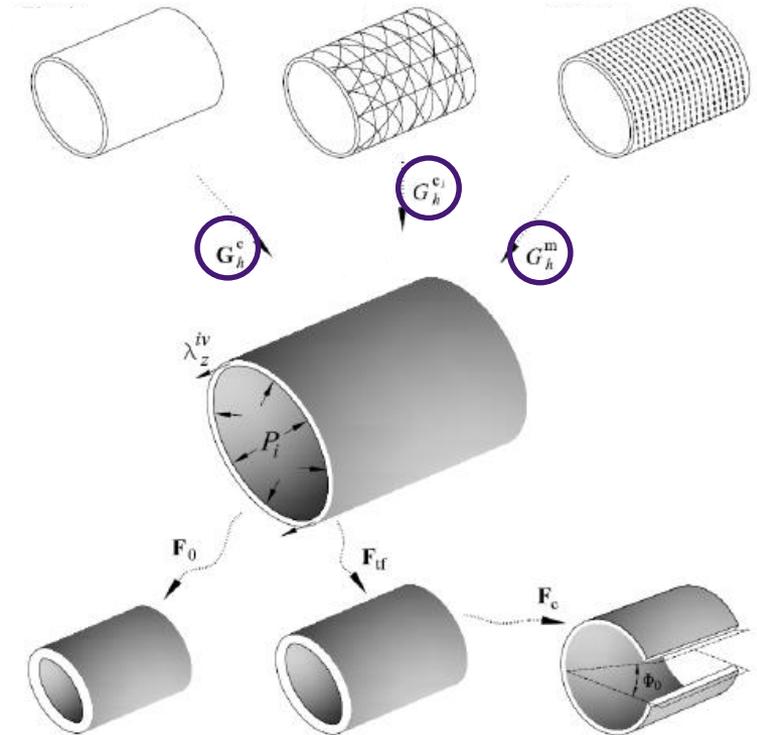
PARAMETERS TO BE IDENTIFIED

$$W = \phi^e W^e(\mathbf{F}^e) + \phi^m W^m(\lambda^m) + \sum_{j=1}^4 \phi^{c_j} W^{c_j}(\lambda^{c_j})$$

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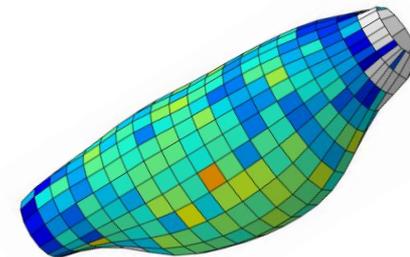
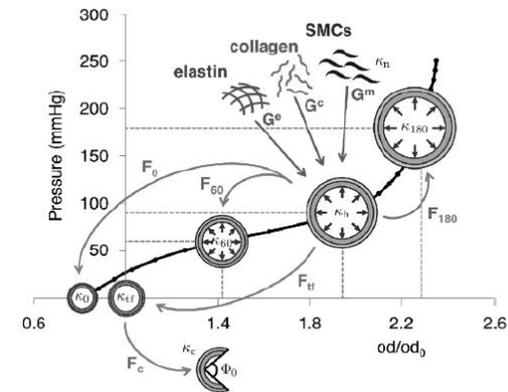
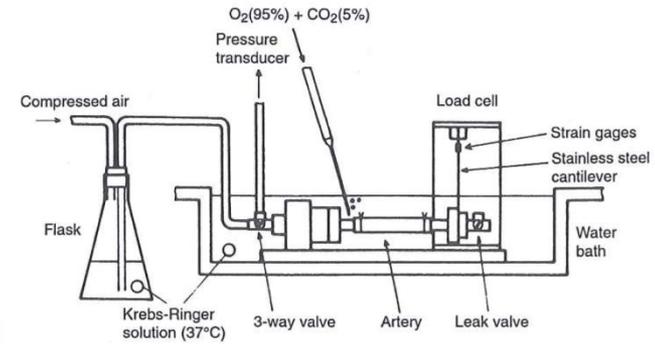
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$$W^c(\lambda^{c_j}) = \frac{c_2^c}{4c_3^c} \left[c_3^c (\lambda^{c_j})^2 - 1 \right]$$

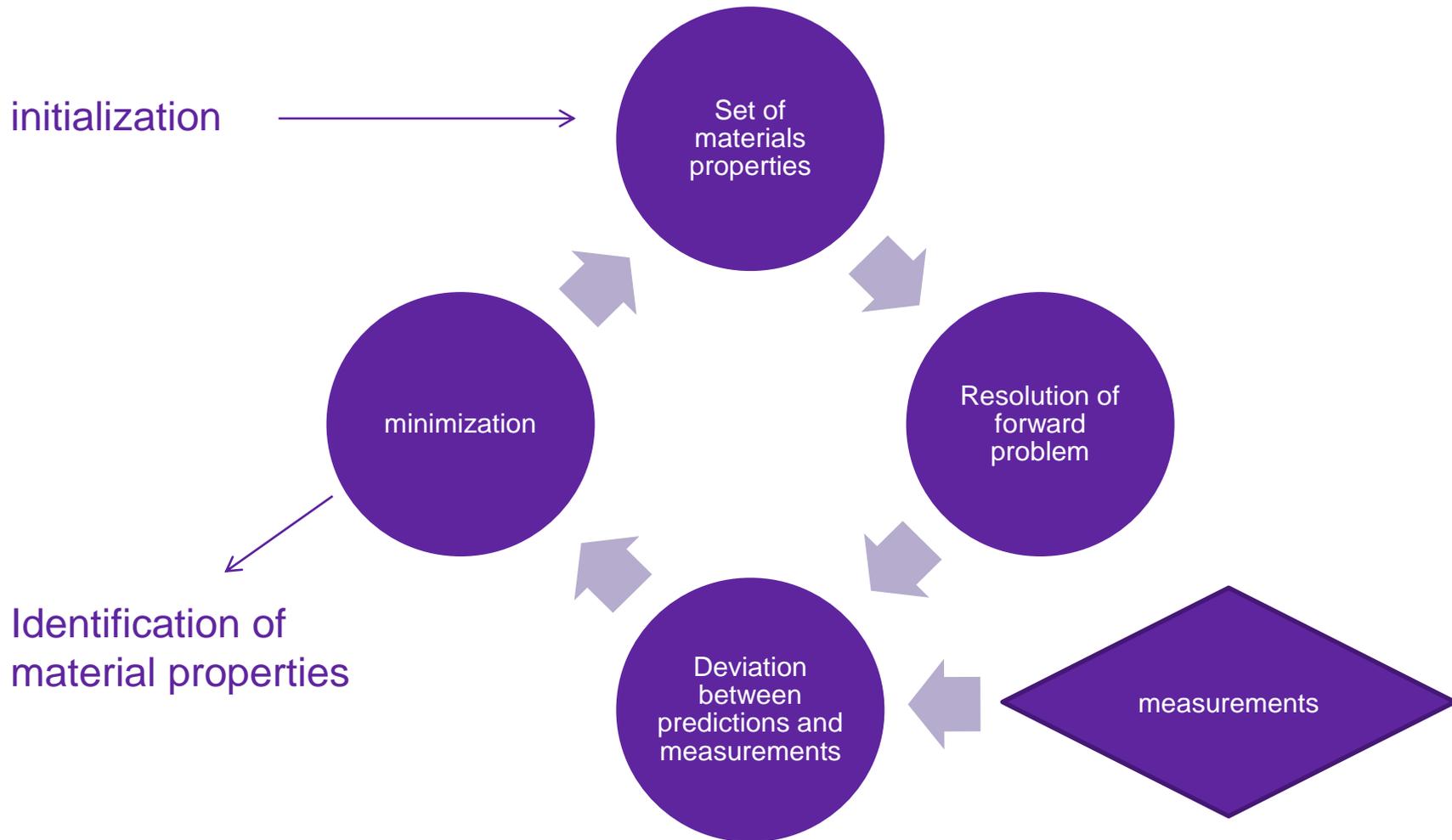


APPROACH

1. Experiments
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Inverse approach – traditional approach

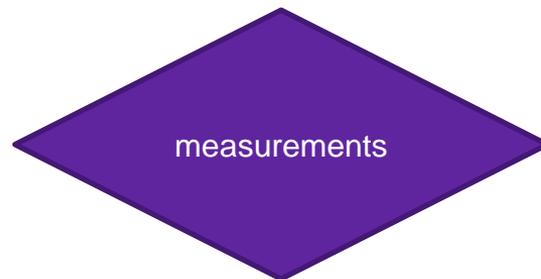
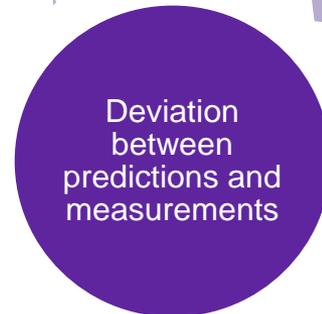
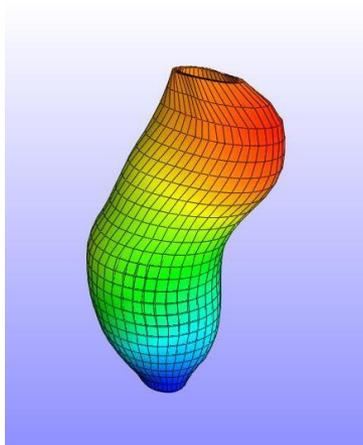


Inverse approach – FEMU approach

initialization

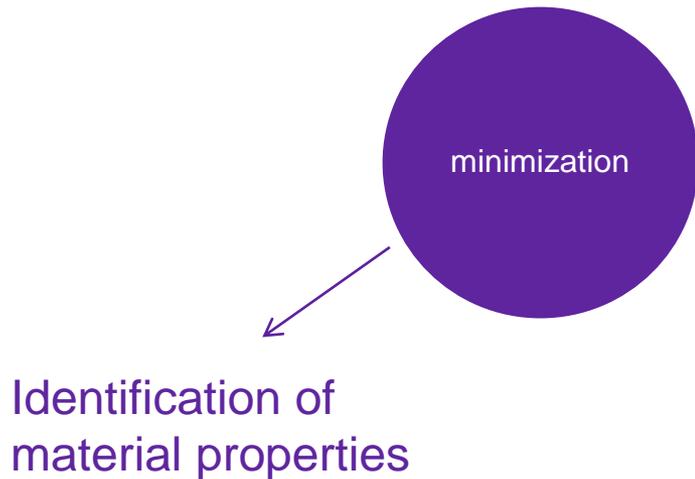


Oberai et al., Inverse problems, **19**, pp. 297-313, 2003



$$J(\mu) = \|T(u) - T(u^{exp})\|^2 + \frac{\alpha}{2} B(\mu)$$

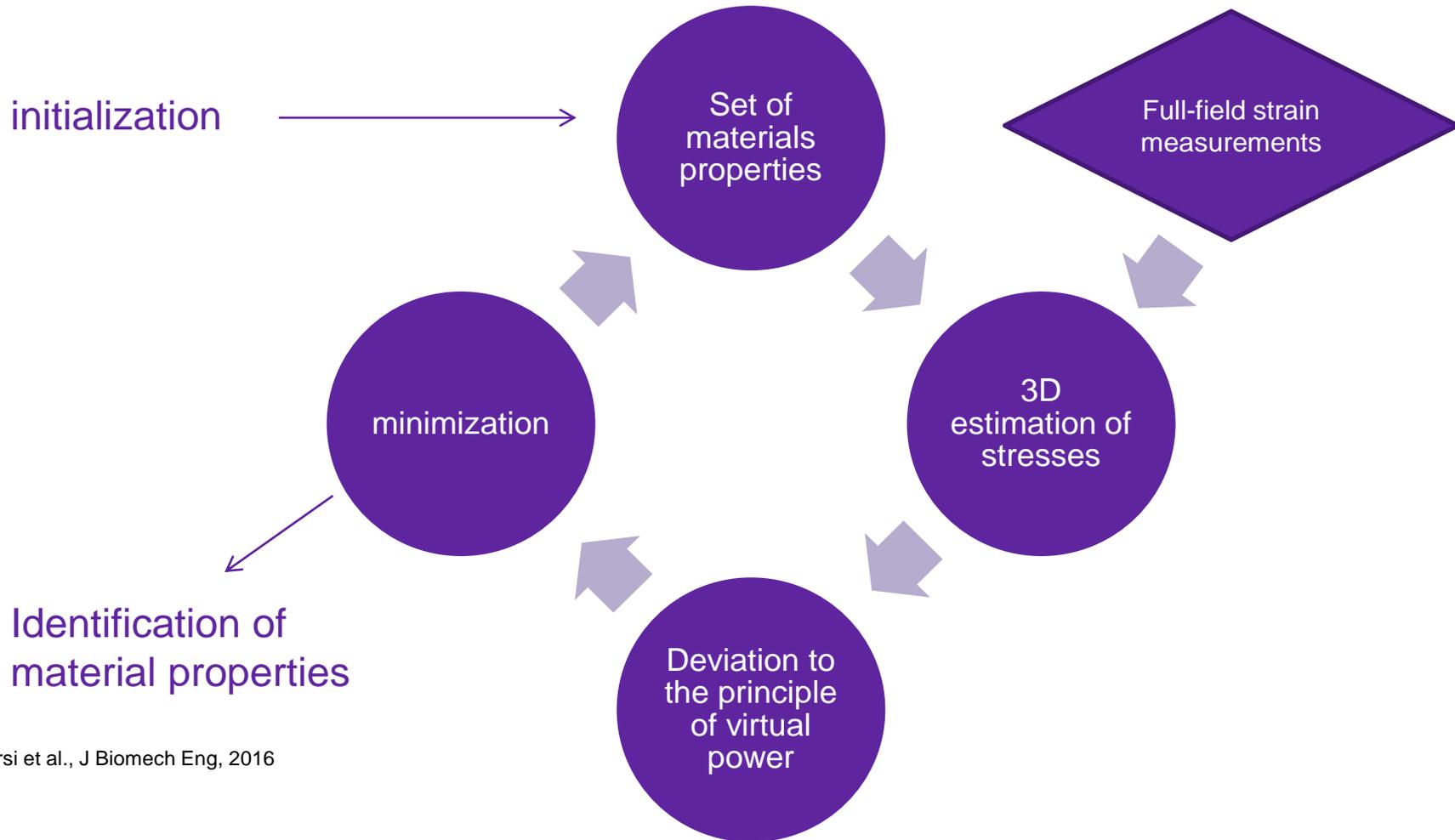
Inverse approach – FEMU approach



1. Use a gradient based method (steepest descent or BFGS)
2. Need to derive the gradient of J with respect to μ at each iteration. With the adjoint method, this requires the resolution of 2 forward problems
3. Very unstable with hyperelastic models: **many risks that the forward problems have a poor convergence**



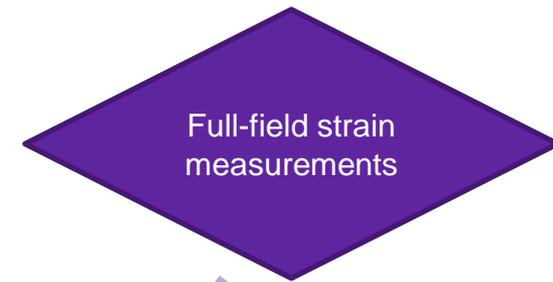
Alternative inverse approach: the virtual fields method



Bersi et al., J Biomech Eng, 2016

Full-field stress reconstruction

initialization



$$W = \phi^e W^e(\mathbf{F}^e) + \phi^m W^m(\lambda^m) + \sum_{j=1}^4 \phi^{c_j} W^{c_j}(\lambda^{c_j})$$

$$W^e(\mathbf{F}^e) = \frac{c^e}{2} \left[\text{tr} \left((\mathbf{F}^e)^T \mathbf{F}^e \right) - 3 \right]$$

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$$W^c(\lambda^{c_j}) = \frac{c_2^c}{4c_3^c} \left[e^{c_3^c ((\lambda^{c_j})^2 - 1)^2} - 1 \right]$$

Simple application of the constitutive model for each element

Minimization of the equilibrium gap using the principle of virtual power

minimization

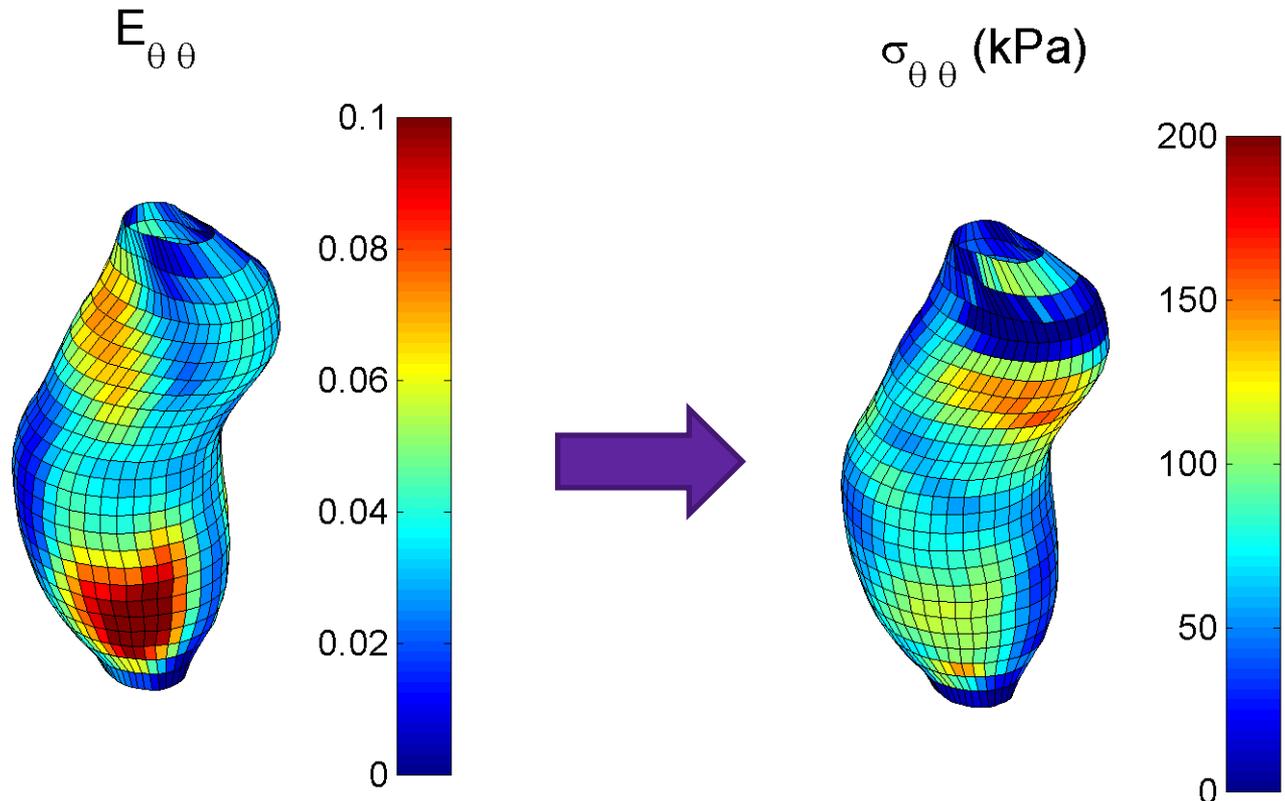
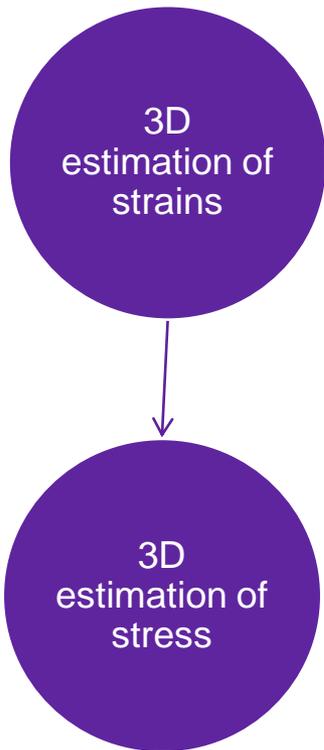
$$J = \sum_p \sum_\lambda \left(\underbrace{- \int_{\omega(t)} \underline{\sigma} : (\underline{\nabla} \otimes \underline{\xi}^*)}_{P_{int}^*} d\omega + \underbrace{\oint_{\partial\omega(t)} \underline{T} : \underline{\xi}^*}_{P_{ext}^*} ds \right)^2$$

Bersi et al., J Biomech Eng, 2016

Resolution:

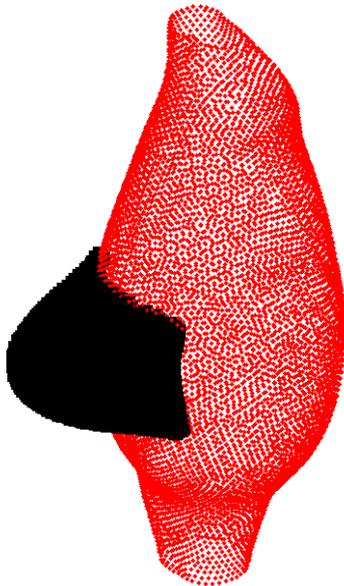
$$\min_{c_3^1, c_3^{2,3}, c_3^4, \alpha, \beta} \left[\underbrace{\min_{c^e, c_2^1, c_2^{2,3}, c_2^4} \left[\frac{J(u)}{A} + \frac{J(v)}{B} \right]}_{\text{Linear least-squares}} \right]_{\text{Genetic algorithm}}$$

Derivation of stress tensor using layer specific constitutive behavior



Virtual power 1

1. A local virtual radial “bulge”: $\mathbf{u}(\mathbf{x}) = [f(\mathbf{x}-\mathbf{x}_0) / r^2] \mathbf{e}_r$



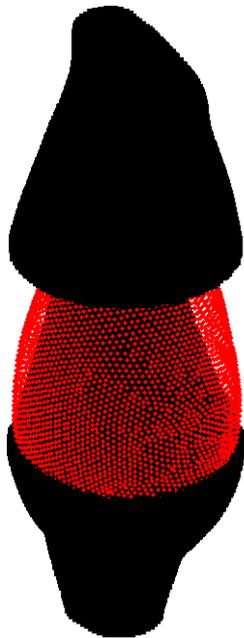
$$\underbrace{- \int_{\omega(t)} \underline{\underline{\sigma}} : \left(\underline{\underline{\nabla}} \otimes \underline{\underline{\xi}}^* \right) d\omega}_{P_{int}^*} + \underbrace{\oint_{\partial\omega(t)} \underline{\underline{T}} : \underline{\underline{\xi}}^* ds}_{P_{ext}^*} = 0$$

After deriving virtual strains (infinitesimal) local internal virtual work is derived at every Gauss point and integrated across the volume

The virtual field is normalized such as the external virtual work equals P.

Virtual power 2

2. Local virtual axial extension: $\mathbf{v}(\mathbf{x}) = f(\mathbf{x}-\mathbf{x}_0) (z \mathbf{e}_z - x/2 \mathbf{e}_x - y/2 \mathbf{e}_y)$



$$\underbrace{- \int_{\omega(t)} \underline{\underline{\sigma}} : \left(\underline{\underline{\nabla}} \otimes \underline{\underline{\xi}}^* \right) d\omega}_{P_{int}^*} + \underbrace{\oint_{\partial\omega(t)} \underline{\underline{T}} : \underline{\underline{\xi}}^* ds}_{P_{ext}^*} = 0$$

After deriving virtual strains (infinitesimal) local internal virtual work is derived at every Gauss point and integrated across the volume

The virtual field is normalized such as the external virtual work equals F.

Minimizing the equilibrium gap

minimization

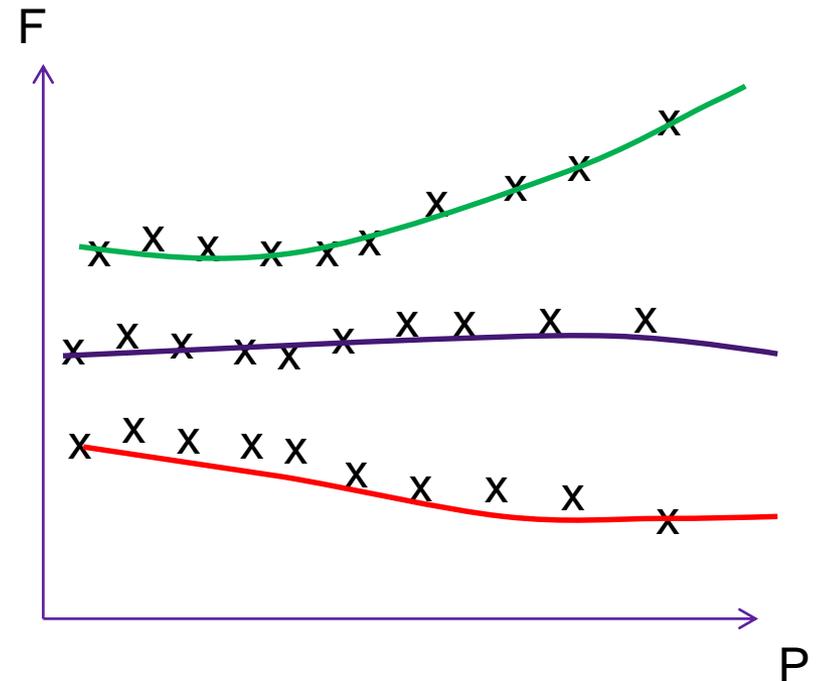
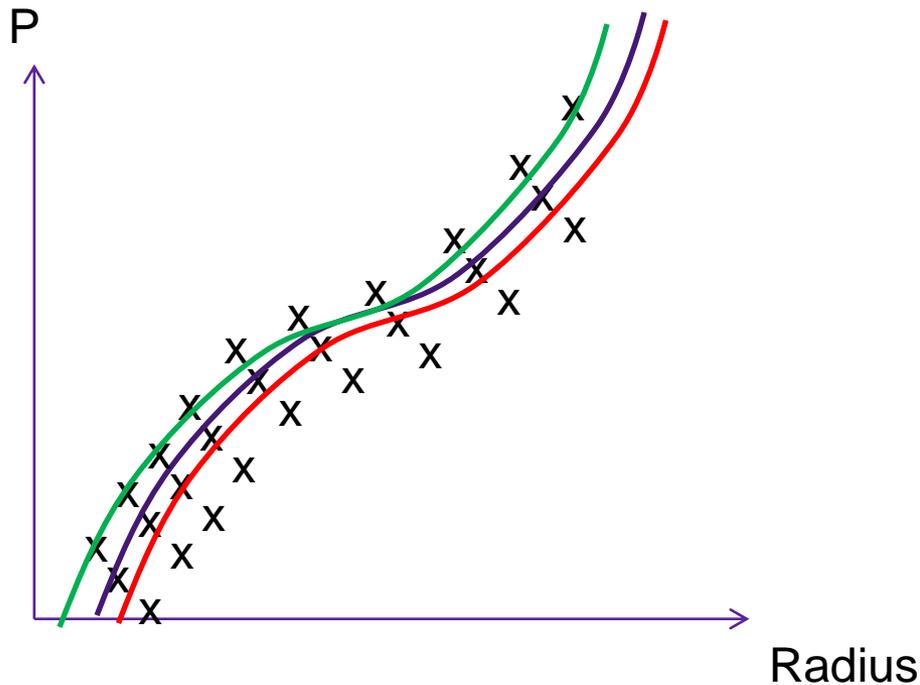
$$J = \sum_p \sum_\lambda \left(\underbrace{- \int_{\omega(t)} \underline{\underline{\sigma}} : \left(\underline{\underline{\nabla}} \otimes \underline{\underline{\xi}}^* \right) d\omega}_{P_{int}^*} + \underbrace{\oint_{\partial\omega(t)} \underline{\underline{T}} : \underline{\underline{\xi}}^* ds}_{P_{ext}^*} \right)^2$$

Bersi et al., J Biomech Eng, 2016

Resolution:

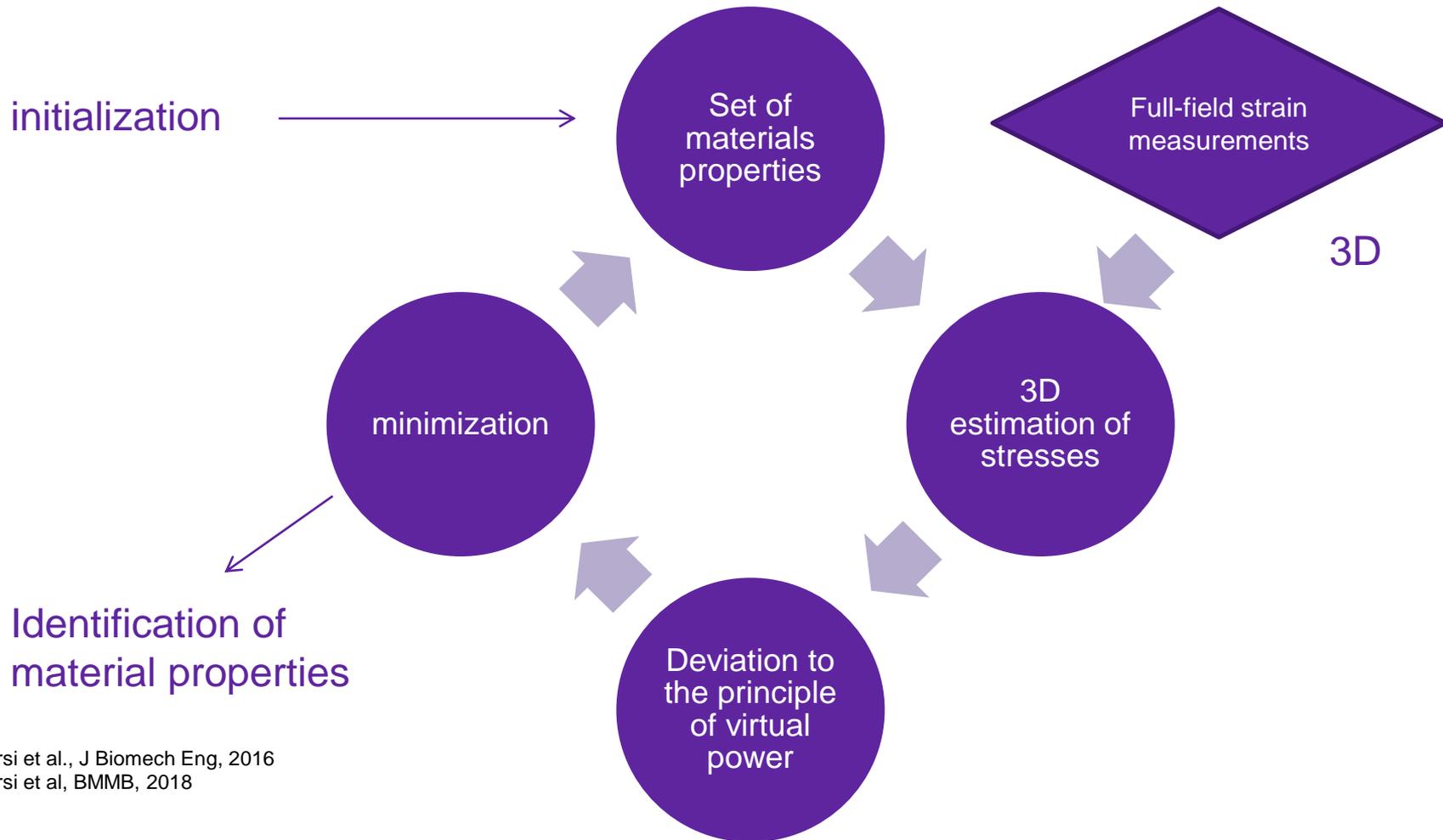
$$\min_{c_3^1, c_3^{2,3}, c_3^4, \alpha, \beta} \left[\underbrace{\min_{c^e, c_2^1, c_2^{2,3}, c_2^4} \left[\frac{J(u)}{A} + \frac{J(v)}{B} \right]}_{\text{Linear least-squares}} \right]_{\text{Genetic algorithm}}$$

Similar to material fitting at every position



Crosses represent external virtual work for every pressure and axial stretch
Solid lines represent internal virtual work
The goodness of fit is evaluated with the R^2 value

Summary of the inverse approach



Bersi et al., J Biomech Eng, 2016
Bersi et al, BMMB, 2018



Results - Highlights

DISSECTED ANEURYSMS



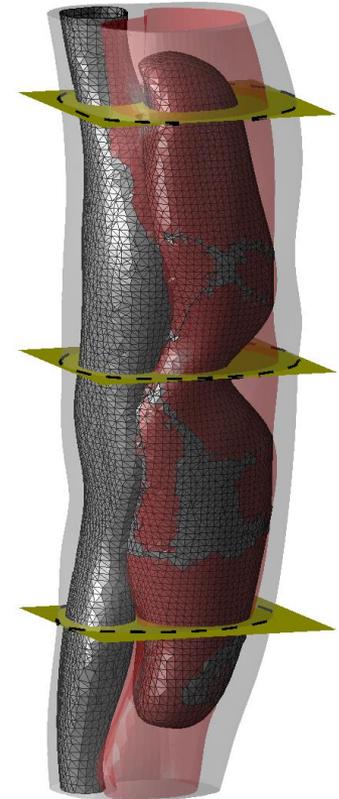
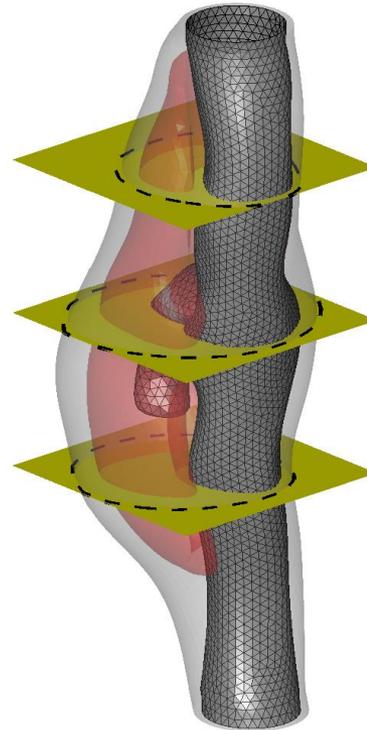
Dissecting Aortic Aneurysm



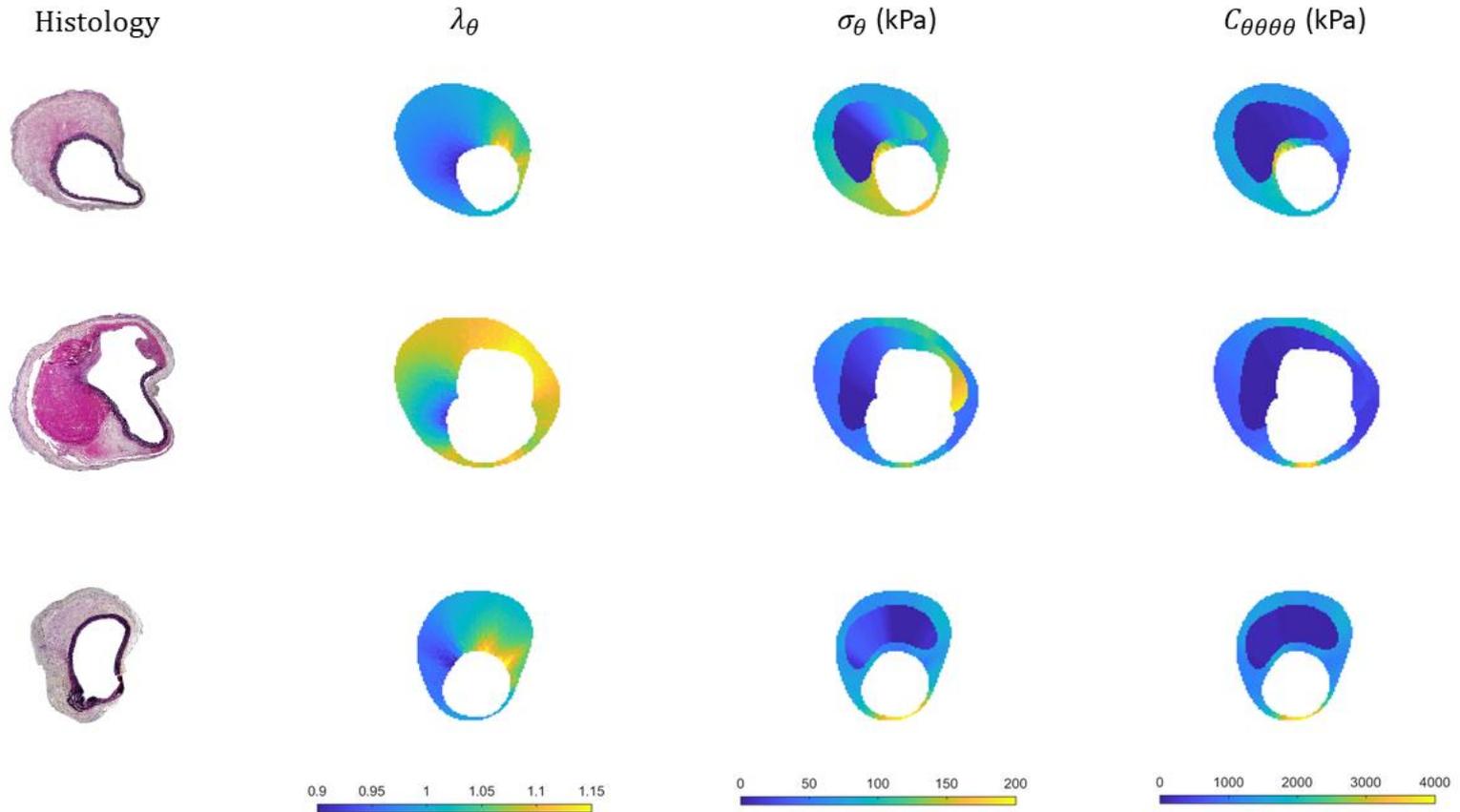
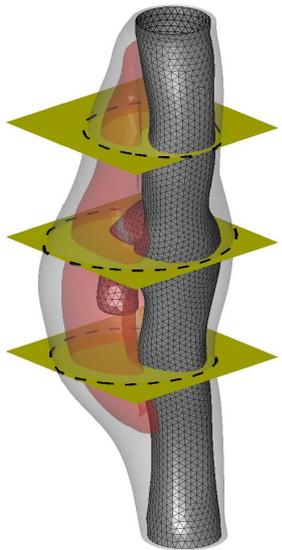
Anterior



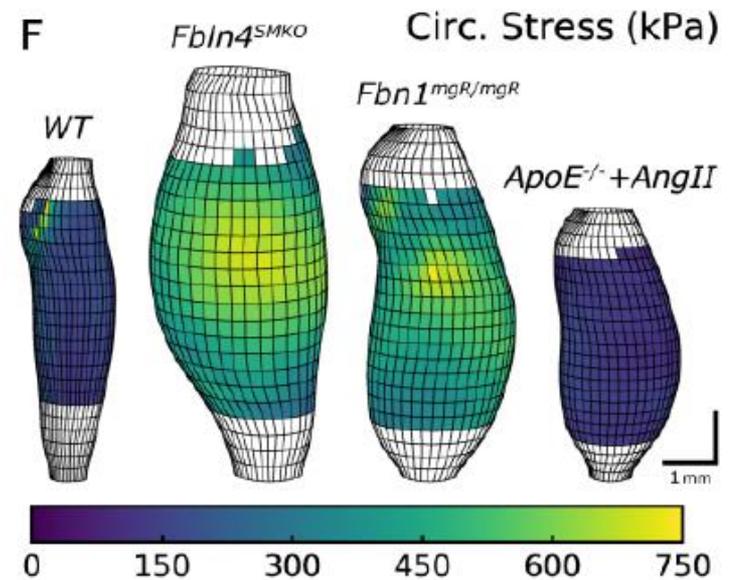
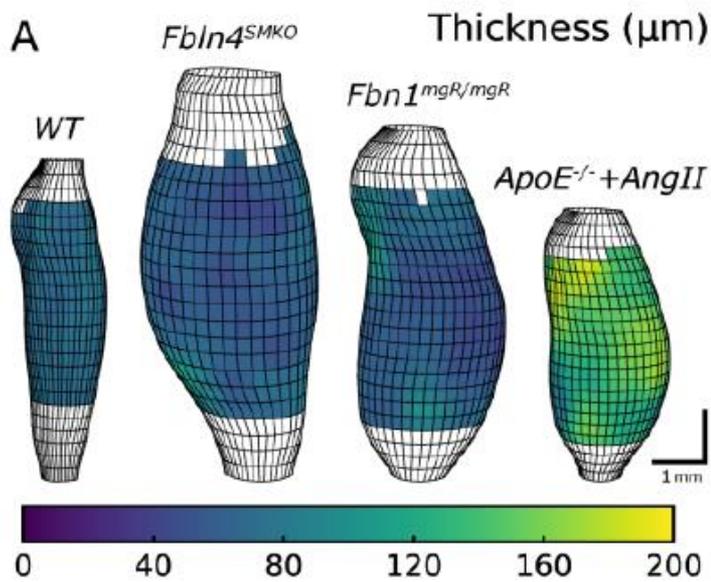
Posterior



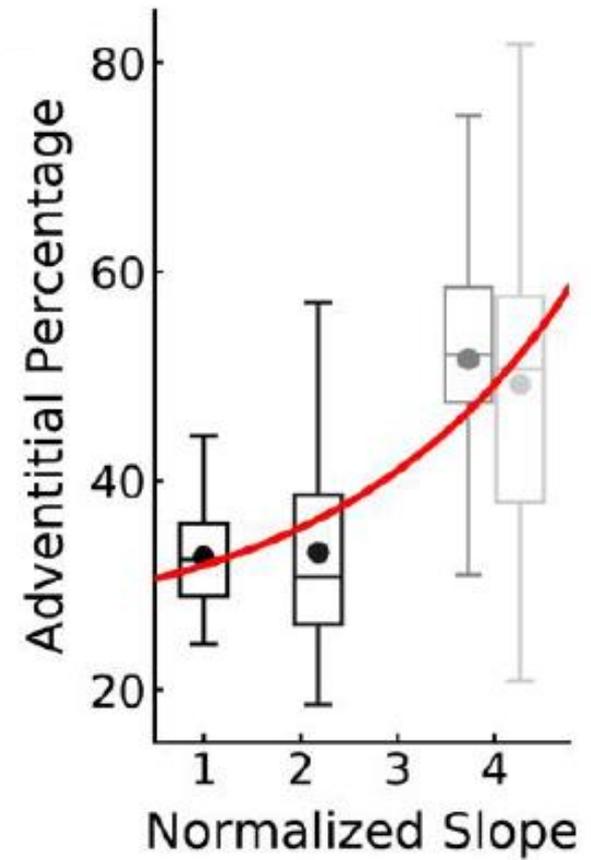
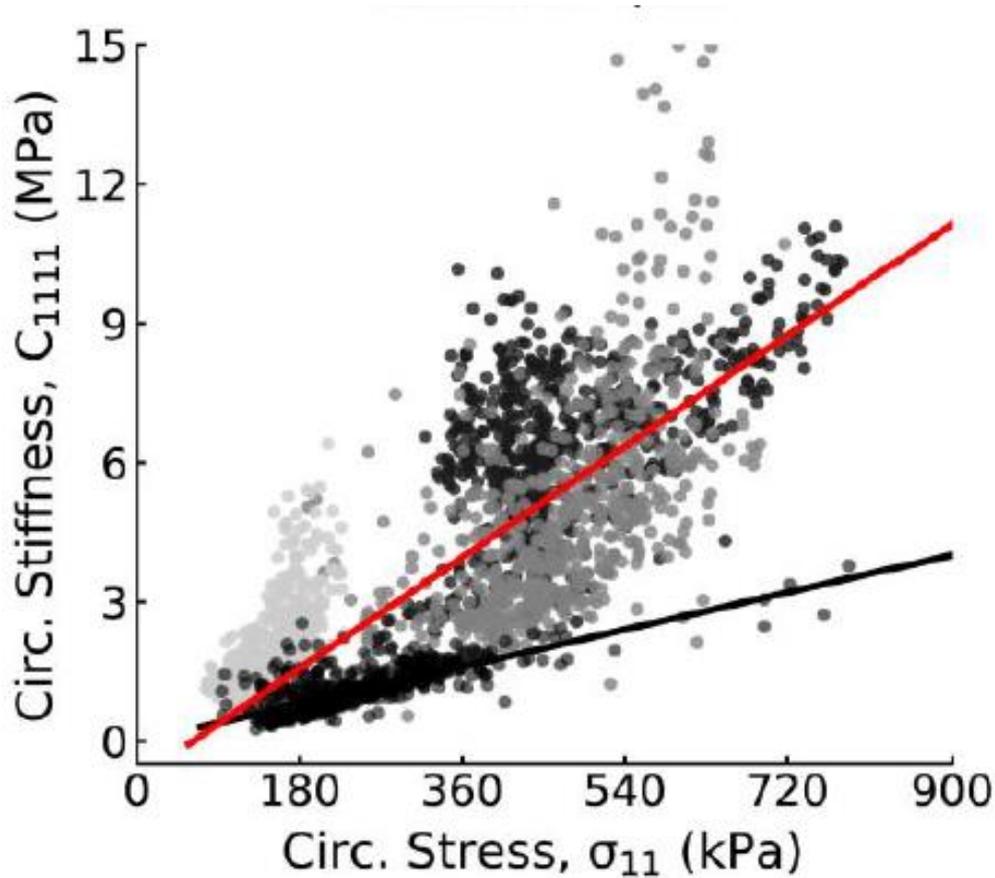
Cross sectional results



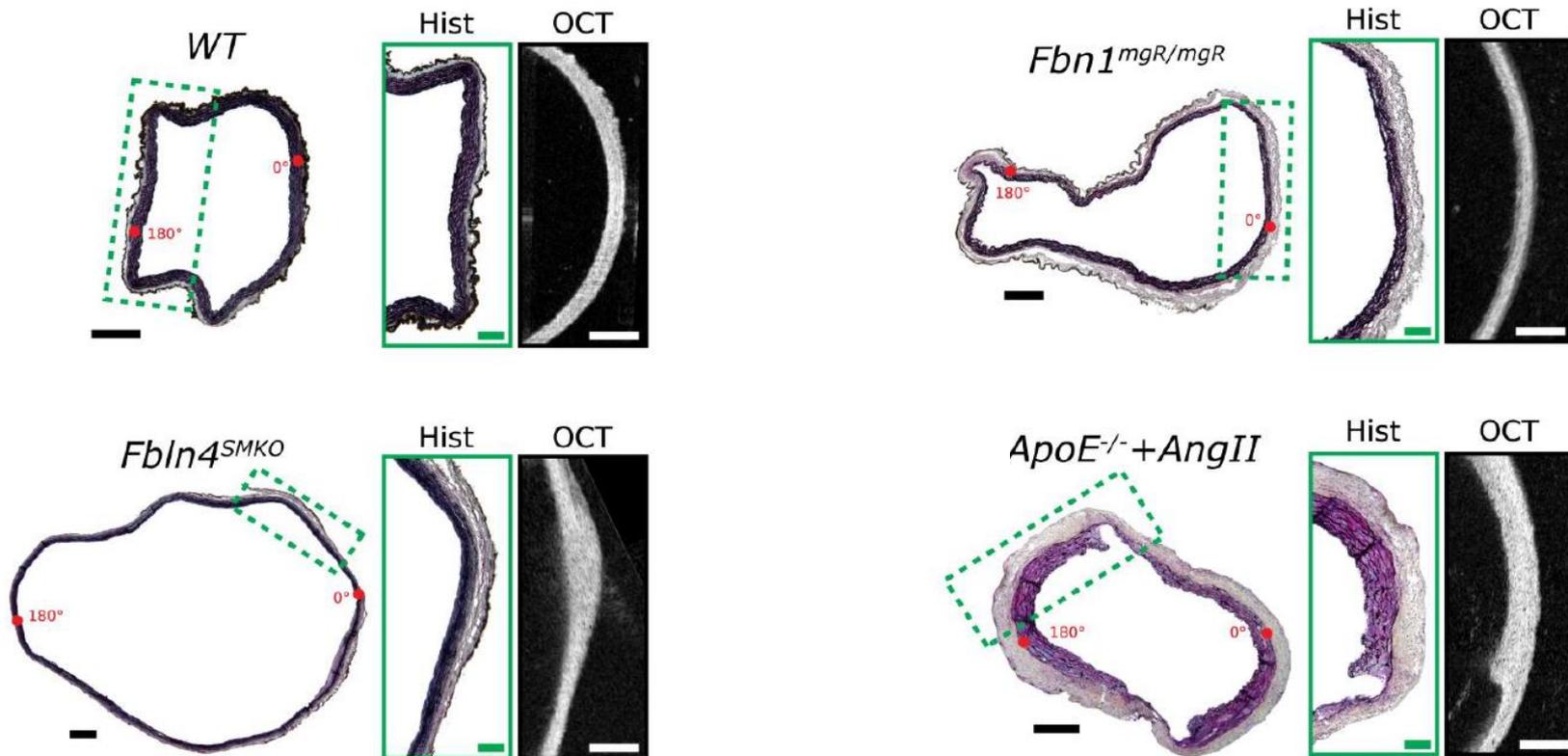
Full-Field Material Parameter Estimation vs thickness distribution



Full-Field Material Parameter Estimation vs local stress

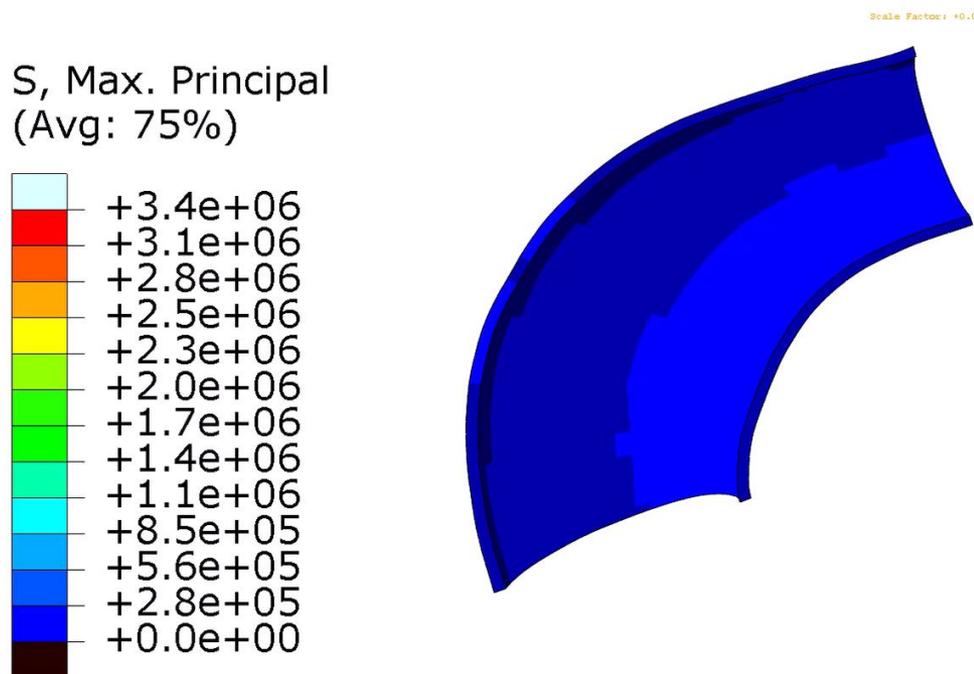


Correlation with tissue μ structure

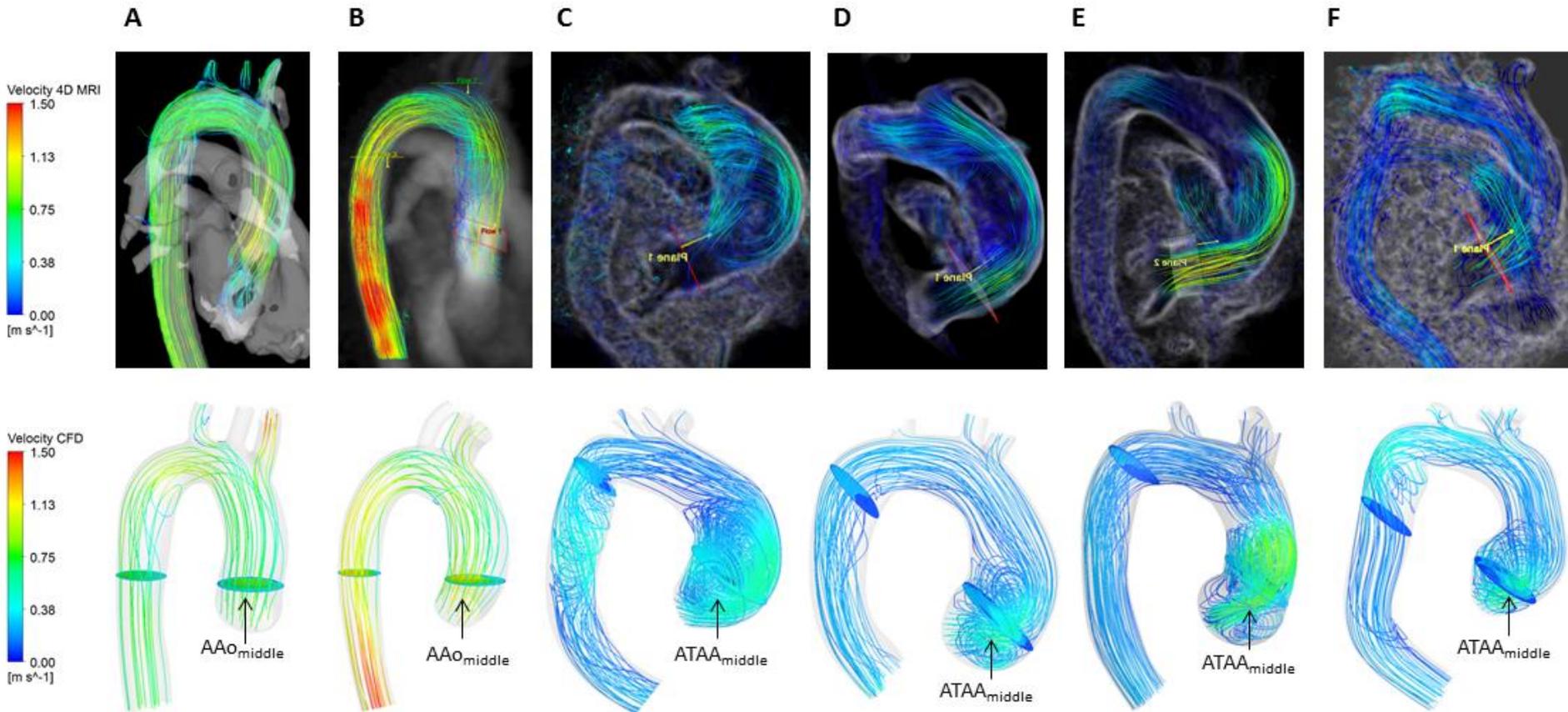


Vision

- Our vision is that the evolution of the strength and of the wall stress of the aorta during the growth of an aneurysm can be predicted on a patient-specific basis by a **computational model**.



Correlation with flow descriptors



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