Efficient Uncertainty Quantification in Patient Specific Assessment of AAA Rupture Risk

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Introduction

AAA rupture risk prediction using FEM

- · Computational rupture risk indicators are superior to the diameter criterion [1]
- Most "patient-specific" models use population averaged model parameters

Existing uncertainties

- · Computational geometries (e.g. wall thickness, stress free configuration)
- Boundary conditions (e.g. intra luminal pressure)
- Physical parameters (e.g. constitutive parameters)

Towards more reliable rupture risk prediction

- In absence of truly patient-specific parameters: Include uncertainties in the FEM analysis
- As a first step in this direction uncertain constitutive parameters are considered

Stochastic Constitutive Law

≥ 13.50

0.23 α [Mpa]

Experimental research

- Tensile tests reveal significant inter- and intra-patient variations
- Random field approach to model fluctuations in the parameter β

Material behavior

 Stochastic extension of Raghavan and Vorp's hyperelastic constitutive model for aneurysmatic arterial wall [2]

$$\Psi(I_1, J, \boldsymbol{x}, \boldsymbol{\xi}) = \alpha(\bar{I}_1 - 3) + \beta(\boldsymbol{x}, \boldsymbol{\xi})(\bar{I}_1 - 3)^2 + \frac{\kappa}{\eta^2}(\eta \ln J + J^{-\eta} - 1)$$

The parameter $\beta(x, \xi)$ is modeled as a three dimensional random field [4]: Marginal probability density:

$$p_{\beta}(x) = \frac{1}{x\sigma\sqrt{2\pi}} e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}}; \ \mu = 1.0857 \ \sigma = 0.9205$$

Autocorrelation function: $R_{\rm NG}(\boldsymbol{\tau}) = \sigma_{\beta}^2 e^{-(\frac{|\boldsymbol{\tau}|}{d})^2}$

Propagation of Uncertainties

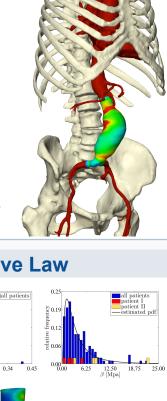
Monte Carlo

· Estimate the distribution of the quantity of interest $\pi_y(y)$ directly using:

$$\pi_y(y) \approx \frac{1}{N_{\text{SAM}}} \sum_{i=1}^{N_{\text{SAM}}} \delta_{y^{(i)}}(y(\boldsymbol{\xi}))$$

- · Minimal implementational overhead
- · Extremely expensive, verification only
- Incorporation of approximate models [3]
- Do sampling on cheap approximate model
- Establish a probabilistic link between high fidelity and approximate model with Bayesian regression

$$\pi(u) = \int p(u|x) \pi(x) dx$$



 $\beta(\boldsymbol{x}_1 + \boldsymbol{ au})$

 $\beta(\boldsymbol{x}_1)$

High fidelity FEM mod

e.g.: coarsening, higher tolerances,

model reduction

Approximate FEM models

Impact on stress

- Overall spatial pattern remains very similar
- · Von Mises stress only mildly depends on β
- Stress state is decoupled from the strain state. Static equilibrium in
- deformed/imaged configuration.
- Prestress determines stress state to a large extend
- AAA is approximately statically determinate

Impact on strains

- Significant impact on spatial strain pattern
- · Very high COV
- Implications for strain or strain energy based damage models [5]

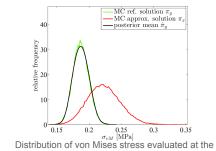
Accuracy

Posterior mean approximation:

$$\hat{\pi}_y(y) = \mathbb{E}_{\boldsymbol{\theta},\sigma}[\pi_y(y)] = \int p(y|x,\boldsymbol{\theta},\sigma^{-2})\pi_x(x)\pi_{\boldsymbol{\theta},\sigma^{-2}}(\boldsymbol{\theta},\boldsymbol{\sigma^{-2}})d\boldsymbol{\theta}d\sigma^{-2}dx$$

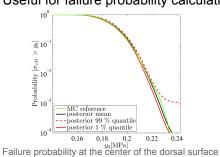
Results

allows accurate prediction of MC reference solution



center of the dorsal surface of the aneurysm sac

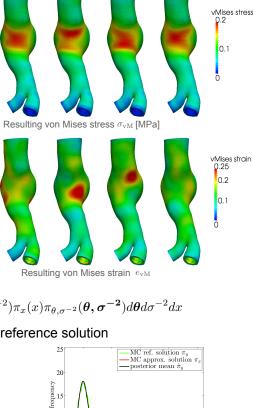
Confidence intervals based on quantiles of the posterior Useful for failure probability calculations

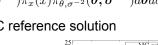


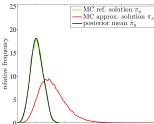
of the aneurysm sac for different failure thresholds

Efficiency

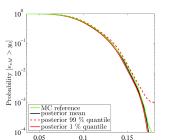
Tremendous reduction in computational costs







0.1 0.2 0.4 $e_{vM}^{0.3}$ 0.5 Distribution of von Mises strain evaluated at the center of the ventral surface of the aneurysm sac

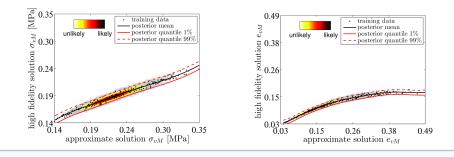


Failure probability at the center of the ventral surface of the aneurysm sac for different failure thresholds

$$p(y|x) \approx p(y|x, \boldsymbol{\theta}, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{(y - f(x, \boldsymbol{\theta}))^2}{2\sigma^2}\right\}$$

- Bayesian regression model $f(x, \theta)$ ٠
- Determination of posterior of the model parameters ٠ using Bayes' rule and advanced SMC scheme with few selected training samples of high fidelity model

 $\frac{p((x_{1:n}, y_{1:n})|\boldsymbol{\theta})p(\boldsymbol{\theta})}{p(x_{1:n}, y_{1:n})} \propto p((x_{1:n}j, y_{1:n})|\boldsymbol{\theta})p(\boldsymbol{\theta})$



- Up to factor 40 cheaper than direct MC on high fidelity model
- Additional potential through numerical continuation schemes

Conclusion and Outlook

- · Population mean values are not good enough for patient-specific assessment of AAA rupture risk
- Strains exhibits large variations whereas stresses are only mildly affected
- Advanced UQ methods cut down the cost to acceptable level
- Include more sources of uncertainty and apply method to a larger patient cohort

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