

Semi-intrusive uncertainty quantification for reliable simulations

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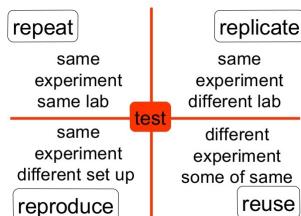
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Credibility

the willingness of persons
to base decisions on information
obtained from the model.^[1]



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Peng RD, Reproducible Research in Computational Science Science 2 Dec 2011: 1226-1227.

^[1] Lee W Schruben. "Establishing the credibility of simulations". In: *Simulation* 34.3 (1980), pp. 101–105.

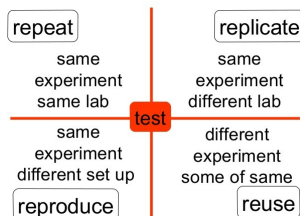
^[2] Lealem Mulugeta et al. "Credibility, replicability, and reproducibility in simulation for biomedicine and clinical applications in neuroscience". In: *Frontiers in neuroinformatics* 12 (2018), p. 18.

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Reproducibility

the ability of a simulation to be copied
by others to provide a simulation
that provides the same results.^[2]



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Credibility

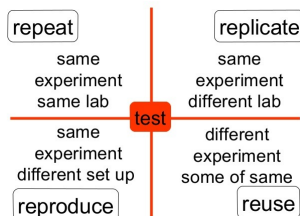
the willingness of persons
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Reliability

the ability of the model to produce accurate and precise results.



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VVUQ as parts of the reliability assessment

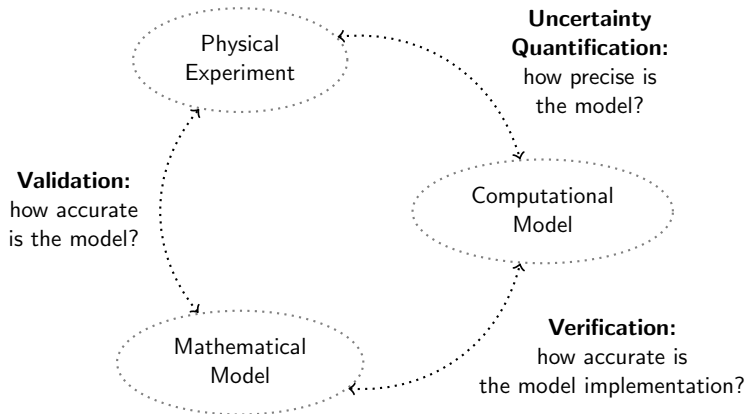


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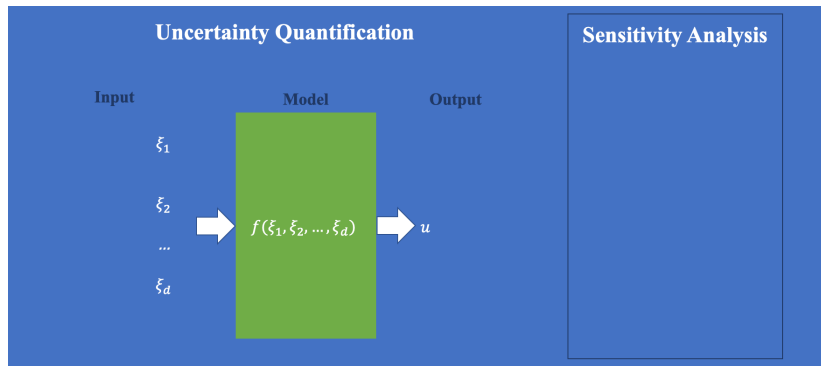
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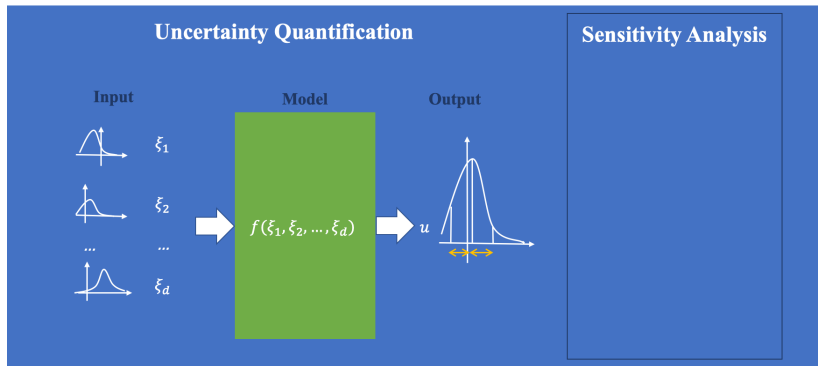
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Uncertainty Quantification (UQ) is a scientific field, which supports decision making using computational models that involve uncertainties.



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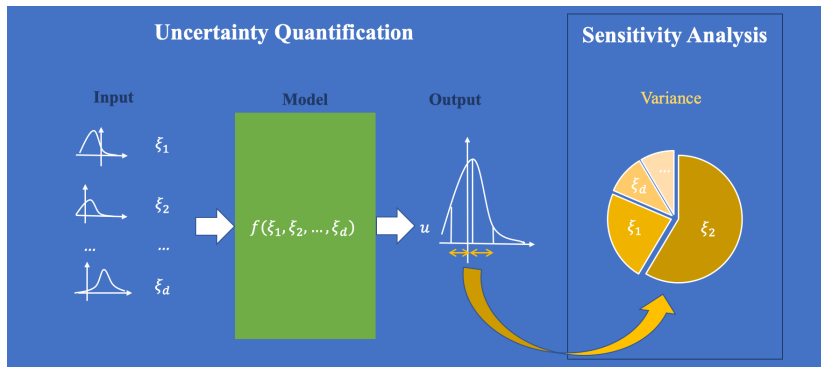
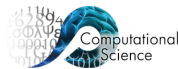


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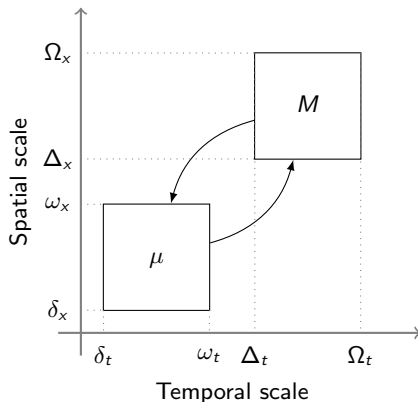
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Scale separation map^[3]



[3] Alfons G Hoekstra et al. "Towards a complex automata framework for multi-scale modeling: Formalism and the scale separation map". In: *International Conference on Computational Science*. Springer. 2007, pp. 922–930.

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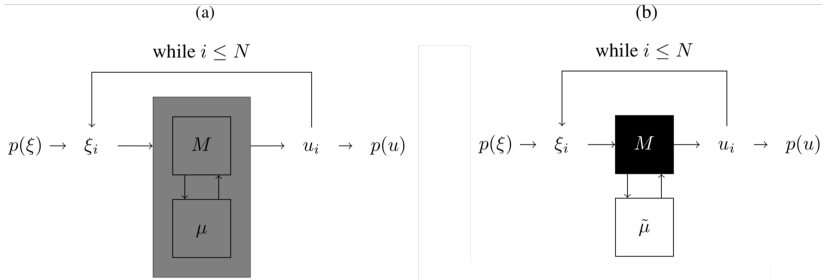
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Non-intrusive vs semi-intrusive methods^[4]



[4] A Nikishova et al. "Semi-intrusive multiscale metamodeling uncertainty quantification with application to a model of in-stent restenosis". In: *Philosophical Transactions of the Royal Society A* 377.2142 (2019), p. 20180154.

Non-intrusive vs semi-intrusive metamodeling methods

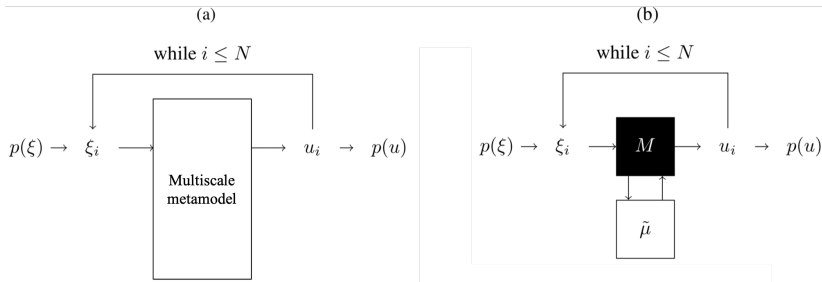


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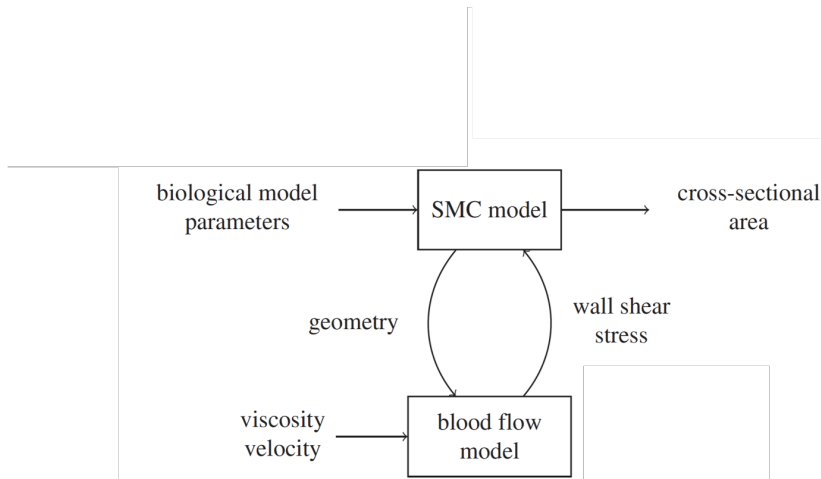
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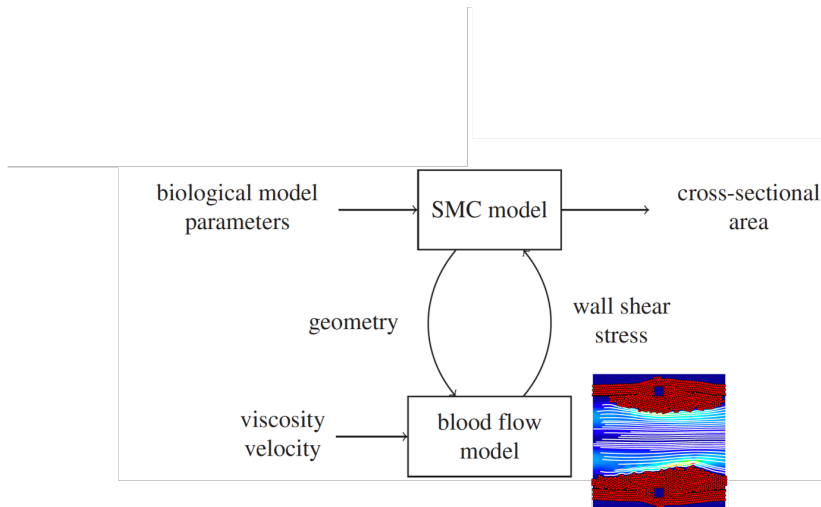
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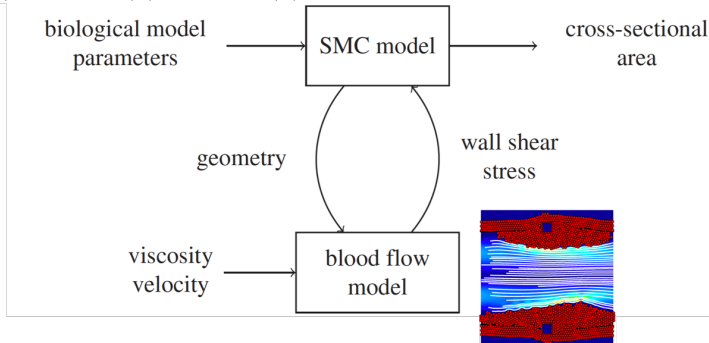
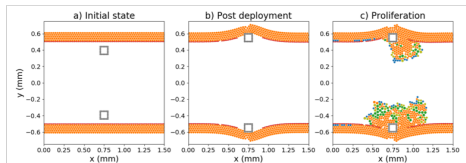
2D multiscale model of in-stent restenosis



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2D multiscale model of in-stent restenosis



UQ methods:

- ▶ **Non-intrusive quasi-Monte Carlo (QMC)**
 - ▶ reference solution
 - ▶ 1024 samples
 - ▶ ~ 90 minutes per sample
- ▶ **Non-intrusive metamodeling (NIM)**
 - ▶ 128 training samples
 - ▶ training takes ~ 5 minutes
 - ▶ ~ 7 times speed up
- ▶ **Semi-intrusive metamodeling (SIUQ)**
 - ▶ 4 training samples
 - ▶ ~ 15 minutes per sample
 - ▶ ~ 7 times speed up

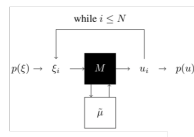
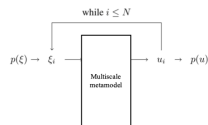
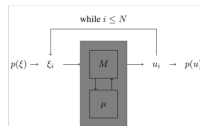


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- ▶ Multiscale models provide advantages in UQ
 - ▶ Better control, since physics is preserved;
 - ▶ More accurate metamodeling;
 - ▶ Analysis of computationally cheap single-scale models.^[5]

[5] Anna Nikishova, Giovanni E Comi, and Alfons G Hoekstra. “Sensitivity analysis based dimension reduction of multiscale models”. In: *Mathematics and Computers in Simulation* 170 (2020), pp. 205–220.

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Thank you!

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