

DELFT UNIVERSITY OF TECHNOLOGY

Project: Error Estimates for Finite Element Simulations Using Neural Networks

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Project Description

Adaptive finite element methods are based on the observation that the error between the discrete finite element solution and the exact solution is not uniformly distributed over the computational domain. For instance, if, due to a loss of regularity, the solution has a singularity, the error is much higher in the vicinity of the singularity; see fig. 1.

It is well known that the accuracy of the discrete solution can be improved by refining the mesh. However, refining the mesh uniformly will reduce the error in all parts of the domain, even in those areas where the error is already sufficiently small. In order to improve the efficiency, the mesh should therefore only be refined in areas with a high error. Since the exact solution and the exact error are typically unknown, a local error estimator is necessary to identify the regions where the mesh has to be refined.

Based on this, an adaptive finite element strategy comprises three components:

- A finite element implementation for a given mesh
- An a posteriori error estimator, which estimates the error of a discrete solution; see, e.g., [2]
- A mesh refinement strategy

The adaptive finite element approach is then essentially a recursive loop in which the mesh is refined locally where the error estimator indicates a high error of the discrete solution. This procedure is repeated until the error estimate is below a certain tolerance. In fig. 1, we can observe that the error can already be reduced significantly by performing two steps of local mesh refinement.

Obviously, the efficiency of the adaptive finite element strategy is particularly determined by the accuracy of the a posteriori error estimator and the mesh refinement strategy. Therefore, the goal of this master project is to investigate if, based on a discrete finite element solution, machine learning techniques, such as neural networks, can be used to indicate which elements should be refined; see [1] for more details on neural networks.

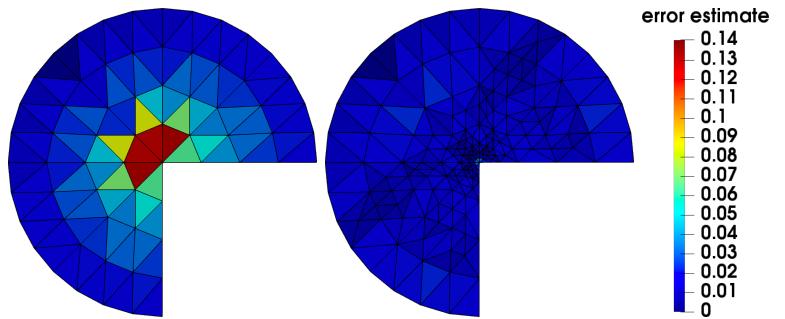


Figure 1: Adaptive mesh refinement on a *PAC-MAN* geometry: residual based a posteriori error estimate on a uniform mesh and after two steps of adaptive mesh refinement.

Tasks

- Install and familiarize with the software packages:
 - The finite element software FEniCS¹.
 - The Python machine learning libraries TensorFlow 2.0² and Keras³.
- Familiarize with adaptive mesh refinement for finite element methods, and implement adaptive finite elements for simple model problems using FEniCS.
- Familiarize with the basic concepts of neural networks.
- Train a neural network in order to predict which elements should be refined in the adaptive finite element implementation.
- Combine the adaptive finite element implementation and the neural network and perform numerical studies to investigate the efficiency of the approach, compared to the initial adaptive finite element implementation.

¹<https://fenicsproject.org>

²<https://www.tensorflow.org>

³<https://keras.io>

Contact

If you are interested in this project and/or have further questions please contact Alexander Heinlein, a.heinlein@tudelft.nl, and Deepesh Toshniwal, d.toshniwal@tudelft.nl.

References

- [1] I. Goodfellow, Y. Bengio, and A. Courville. *Deep learning*. MIT press, 2016.
- [2] R. Verfürth. *A Posteriori Error Estimation Techniques for Finite Element Methods*. Numerical Mathematics and Scientific Computation. OUP Oxford, 2013.