

### DNN Surrogate for Solving PDES with Spatially Varying Coefficients

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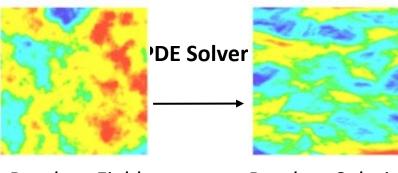
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## Motivation



 PDE with spatially varying but uncertain coefficients. Want to characterize uncertainty in the output using (e.g.) MC methods.

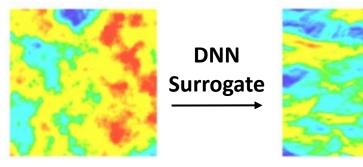


Random Field

Random Solution

#### Drawbacks:

- The PDE solver is very expensive (esp. if input is high dimensional)
- Need thousands of forward solves to obtain convergent statistics



Random Field

Random Solution

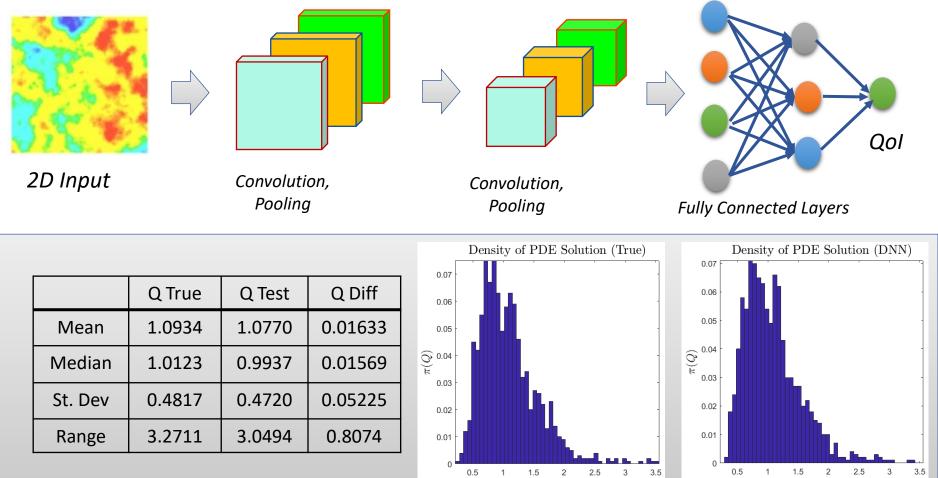
### **Benefits:**

- Trained surrogate is cheaper to evaluate
- Outputs from the surrogate have same distribution as outputs from PDE solver

# Methods & Preliminary Results

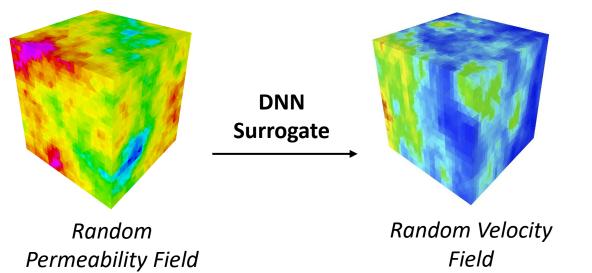


• Trained a convolutional neural network with fully connected layers





## Future Work



- Apply the surrogate to problems with 3D random-field inputs
- Utilize DNN surrogates to accelerate Bayesian inference

#### **References:**

- X. Lou and A. Kareem *Deep convolutional neural networks for uncertainty propagation in random fields*. Computer Aided Civil and Infrastructure Engineering (2019)
- **R.K. Tripathy and I. Bilionis** *Deep UQ: Learning deep neural network surrogate models for high dimensional uncertainty quantification.* Journal of Computational Physics (2018) 565-588.

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