

Problem definition (physics + discretization)

<u>Category</u>	<u>Setting / Value</u>
<i>Problem type</i>	<i>2D plane-strain Neo-Hookean elasticity (square domain)</i>
<i>Domain</i>	<i>$([0, L] \times [0, L])$, with $(L = 1.0)$</i>
<i>Mesh type</i>	<i>Structured quadrilateral mesh (2D square)</i>
<i>Mesh resolution</i>	<i>$(N = 42)$ (42×42 grid; used by <code>create_2D_square_mesh(L, N)</code>)</i>
<i>FE element</i>	<i>4-node quad, Neo-Hookean (<code>NeoHookeMechanica1Loss2DQuad</code>)</i>
<i>Gauss integration</i>	<i><code>num_gp = 2</code> (2×2 Gauss points per element)</i>
<i>Material model</i>	<i>Neo-Hookean, isotropic</i>
<i>Young's modulus</i>	<i>$(E = 1.0)$</i>
<i>Poisson's ratio</i>	<i>$(\nu = 0.3)$</i>
<i>BC type</i>	<i>Dirichlet on left/right boundaries for (U_x, U_y)</i>
<i>BC – left edge</i>	<i>$(U_x = 0.0), (U_y = 0.0)$</i>
<i>BC – right edge</i>	<i>$(U_x = 0.1), (U_y = 0.1)$</i>
<i>Control type</i>	<i>FourierControl (spatially varying stiffness field $(K(x,y))$)</i>
<i>Fourier x-frequencies</i>	<i>$([2, 4, 6])$</i>
<i>Fourier y-frequencies</i>	<i>$([2, 4, 6])$</i>

Category

Setting / Value

Fourier z-frequencies

([0]) (effectively 2D)

Fourier amplitude scale

beta = 20

K field range

min = 1e-1, max = 1.0 (clamp of heterogeneity field)

Samples in dictionary

200 Fourier coefficient vectors (when generated via create_random_fourier_samples)

PI-FNO model & training hyperparameters (parametric/otf/transfer)

Category

Setting / Value

Operator model

FourierNeuralOperator2D

FNO modes (x)

modes1 = 6

FNO modes (y)

modes2 = 6

Channel width

width = 8

Network depth

depth = 4 residual blocks

Last projection channels

channels_last_proj = 32

Output channels

out_channels = 2 (displacement (u_x, u_y))

<u>Category</u>	<u>Setting / Value</u>
Output scaling	<code>output_scale = 0.1</code>
Total number of params	$\approx 37,538$ (printed in log)
FOL wrapper	<code>PhysicsInformedFourierParametricOperatorLearning</code>
Loss	Physics-informed Neo-Hookean FE residual (<code>NeoHookeMechanicalLoss2DQuad</code>)
Optimizer	Optax Adam, learning rate <code>1e-3</code>
LR scheduler (unused here)	<code>linear_schedule(1e-4 \rightarrow 1e-5, transition_steps = 5000)</code>
Max epochs	<code>num_epochs = 5000</code>
Convergence settings	<code>relative_error = 1e-100, absolute_error = 1e-100</code> (effectively “epoch-based”)
Train batch size (param.)	<code>batch_size = 5</code>
Train indices	<code>0-179 \rightarrow 180 samples</code>
Test indices	<code>181-199 \rightarrow 19 samples</code> (sample 180 reserved for OTF)
Training plots	<code>total_loss, residual_rms_batch_mean</code> (physics residual RMS)
Train checkpoint	Least-loss checkpoint every 100 epochs \rightarrow <code>.../flax_train_state</code>
Final state save	<code>flax_final_state</code> in the parametric output folder

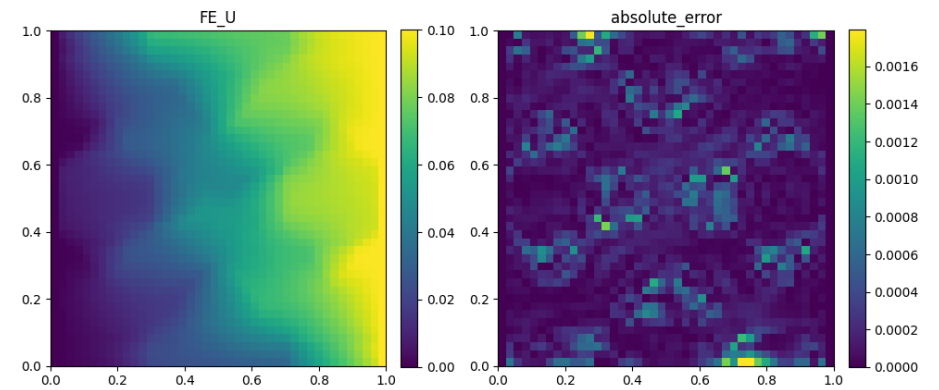
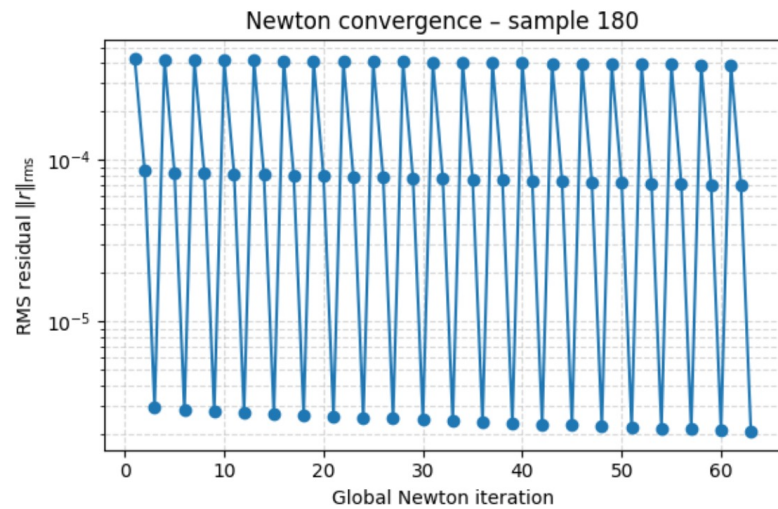
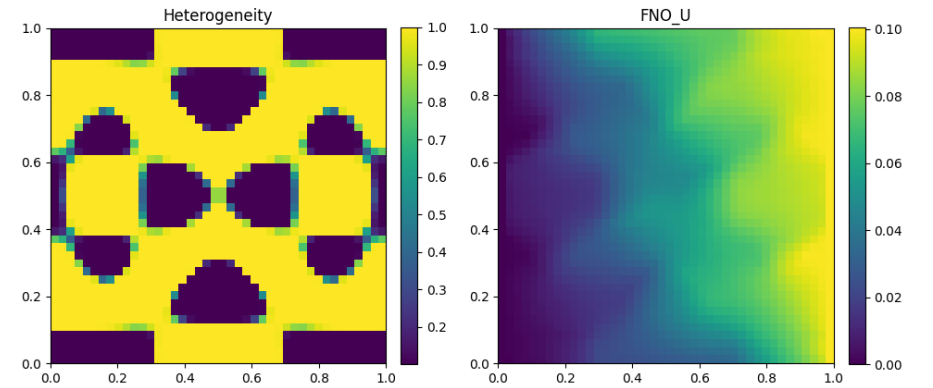
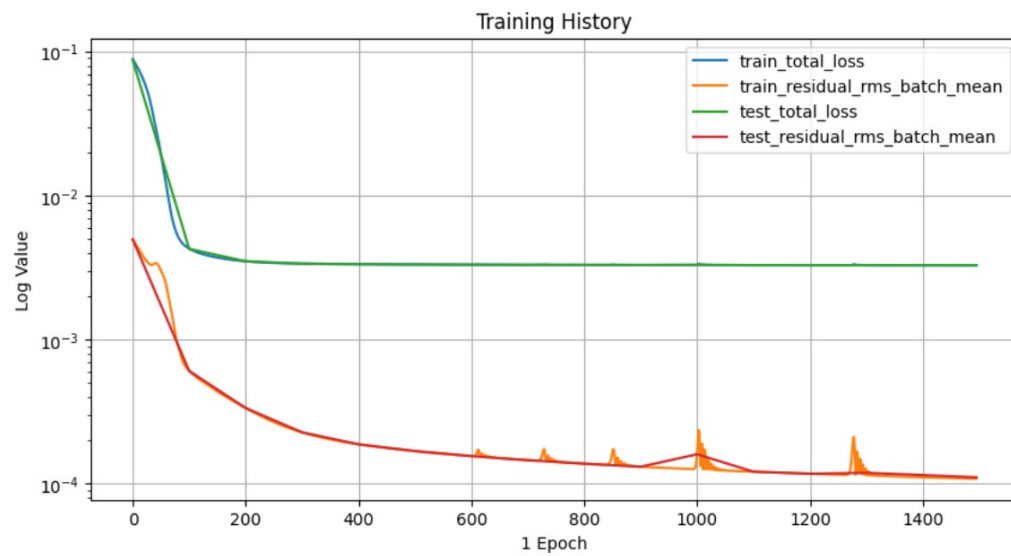
OTF fine-tuning & FE solver hyperparameters (slide 3: “OTF from param”)

OTF PI-FNO (starting from parametric weights)

Category	Setting / Value
<i>Initial weights</i>	<i>Restored from parametric run:</i> <i>./nn_output_mechanical_2D_neohooke_pi_fno_param/flax_train_state</i>
<i>OTF sample ID</i>	<i>otf_id = 180</i>
<i>Train set (OTF)</i>	<i>Only coefficient vector of sample 180 (shape $(1 \times n_{\text{coeffs}})$)</i>
<i>Test set (OTF)</i>	<i>Same as train: sample 180</i>
<i>parametric_learning flag</i>	<i>False (enables true on-the-fly mode)</i>
<i>Eval cases</i>	<i>eval_cases = [180]</i>
<i>Batch size (OTF)</i>	<i>batch_size = 1</i>
<i>Max epochs (OTF)</i>	<i>num_epochs = 5000</i>
<i>Convergence settings</i>	<i>Same: relative_error = 1e-100, absolute_error = 1e-100</i>
<i>Plots during OTF</i>	<i>total_loss, residual_rms_batch_mean saved in</i> <i>nn_output_mechanical_2D_neohooke_pi_fno_otf_from_param</i>
<i>Checkpoint directory (OTF)</i>	<i>./nn_output_mechanical_2D_neohooke_pi_fno_otf_from_param/flax_train_state</i>

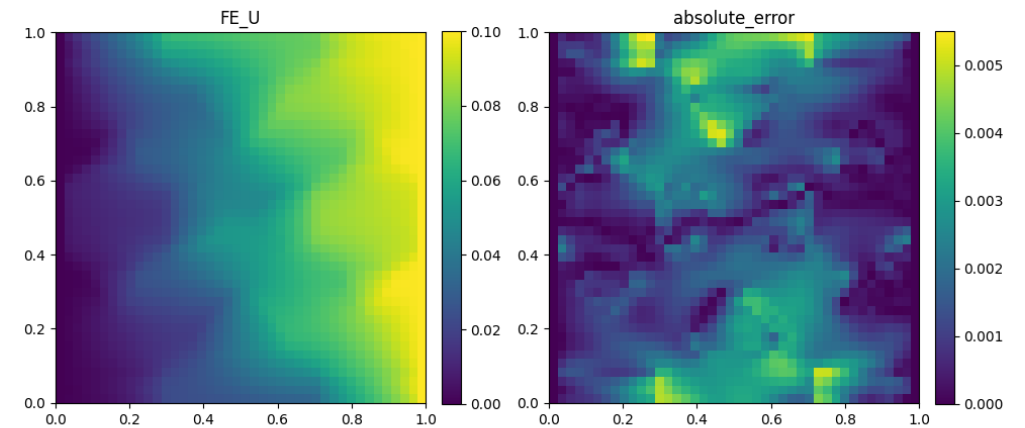
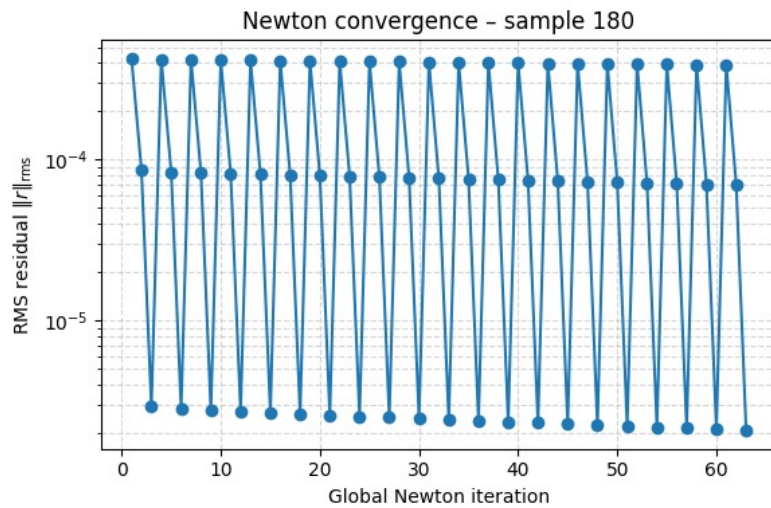
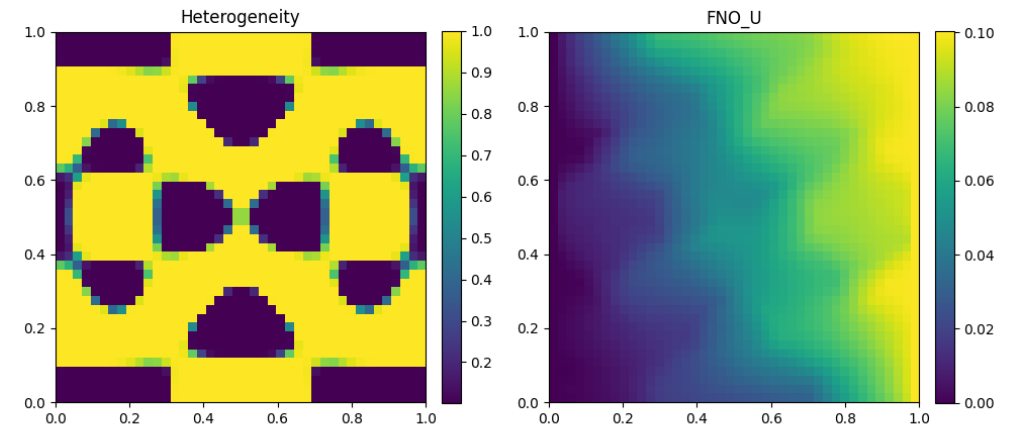
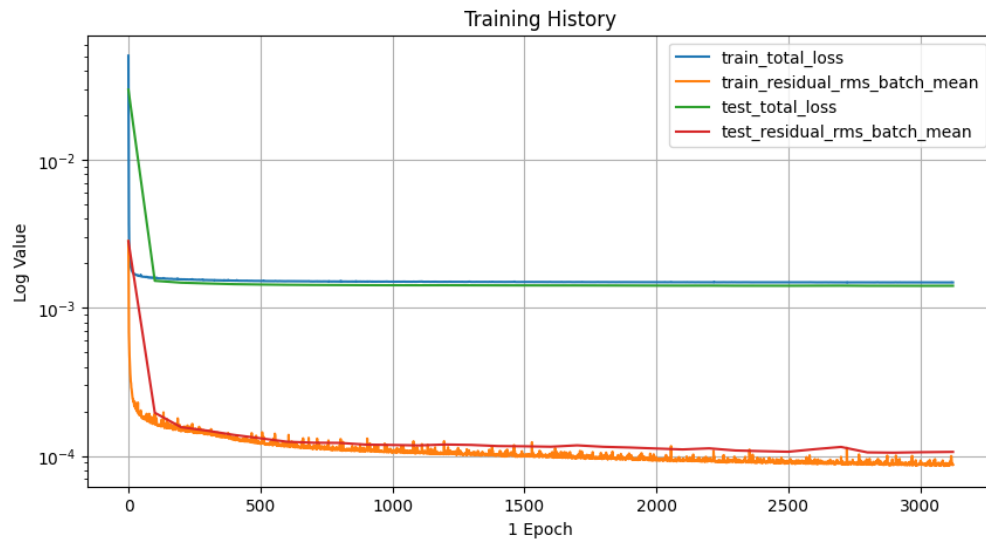
OTF- SLIDE 1

10 mins for training -2000 epochs

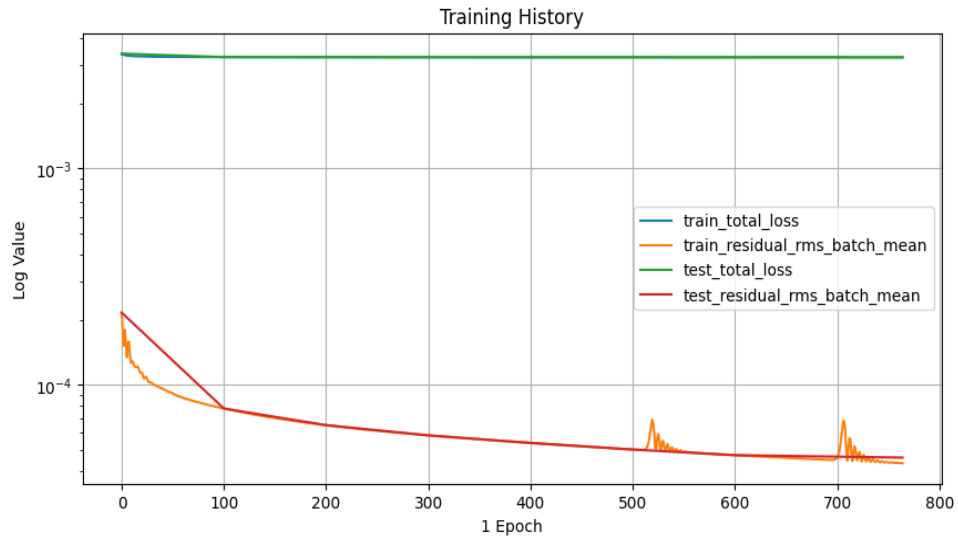


PARAMETRIC- SLIDE 2

30 mins for training – Auto-Stopped between 3000-3500 epochs



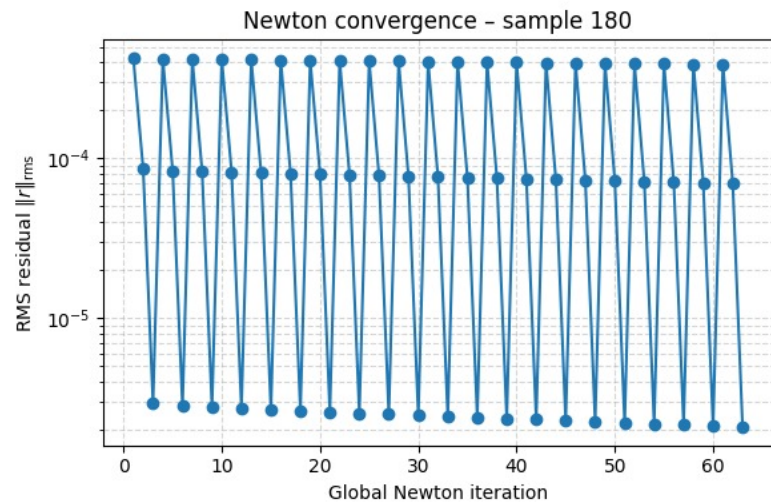
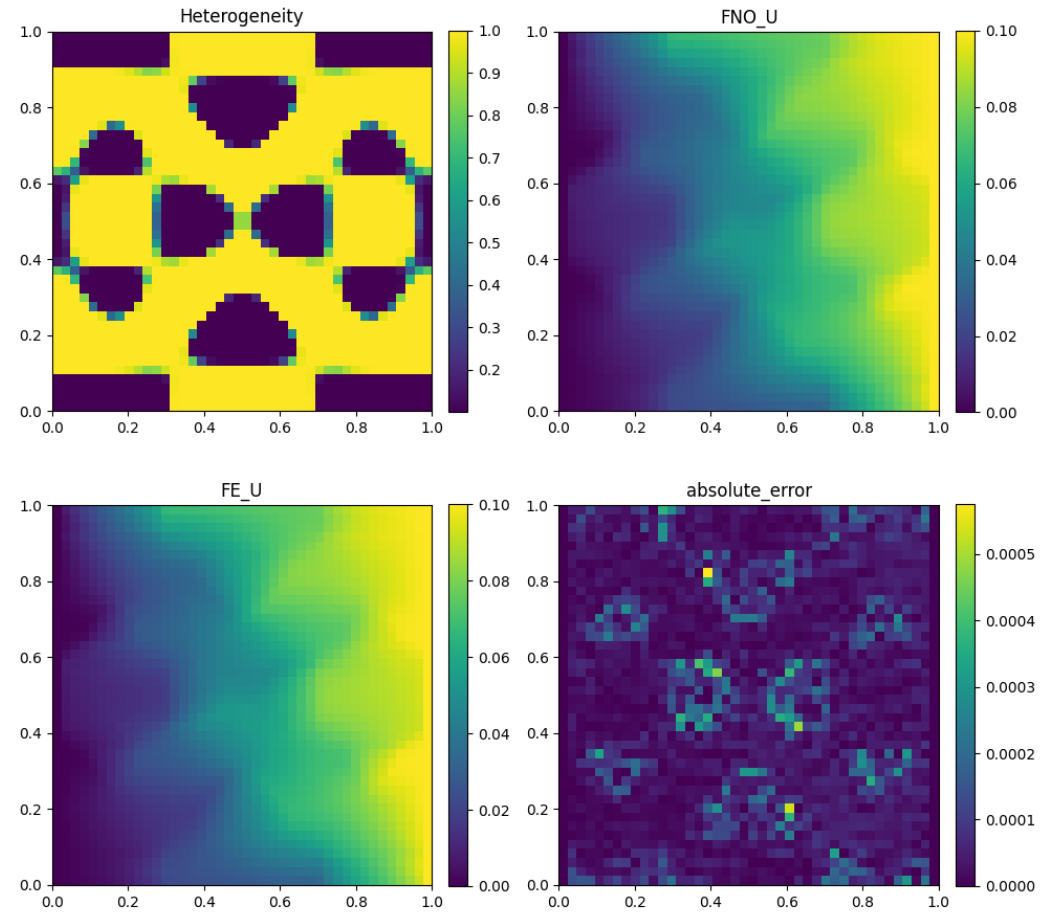
PARAMETRIC TRANSFER LEARN- SLIDE 3.1



5 mins for training — Auto-Stopped between 700-800 epochs

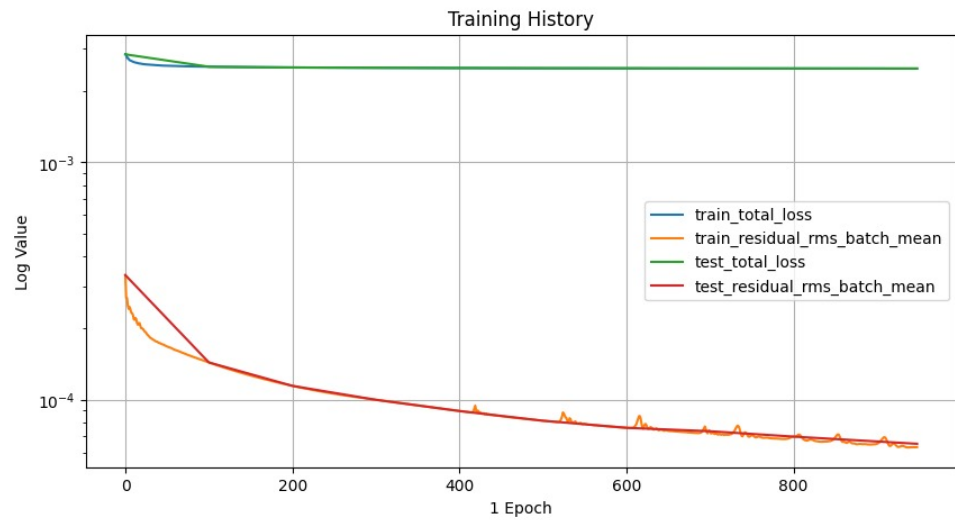
Absolute Error reduction = $0.005 - 0.0005 = 0.0045$

Relative improvement = $0.0045 / 0.005 = 0.9 = 90\%$

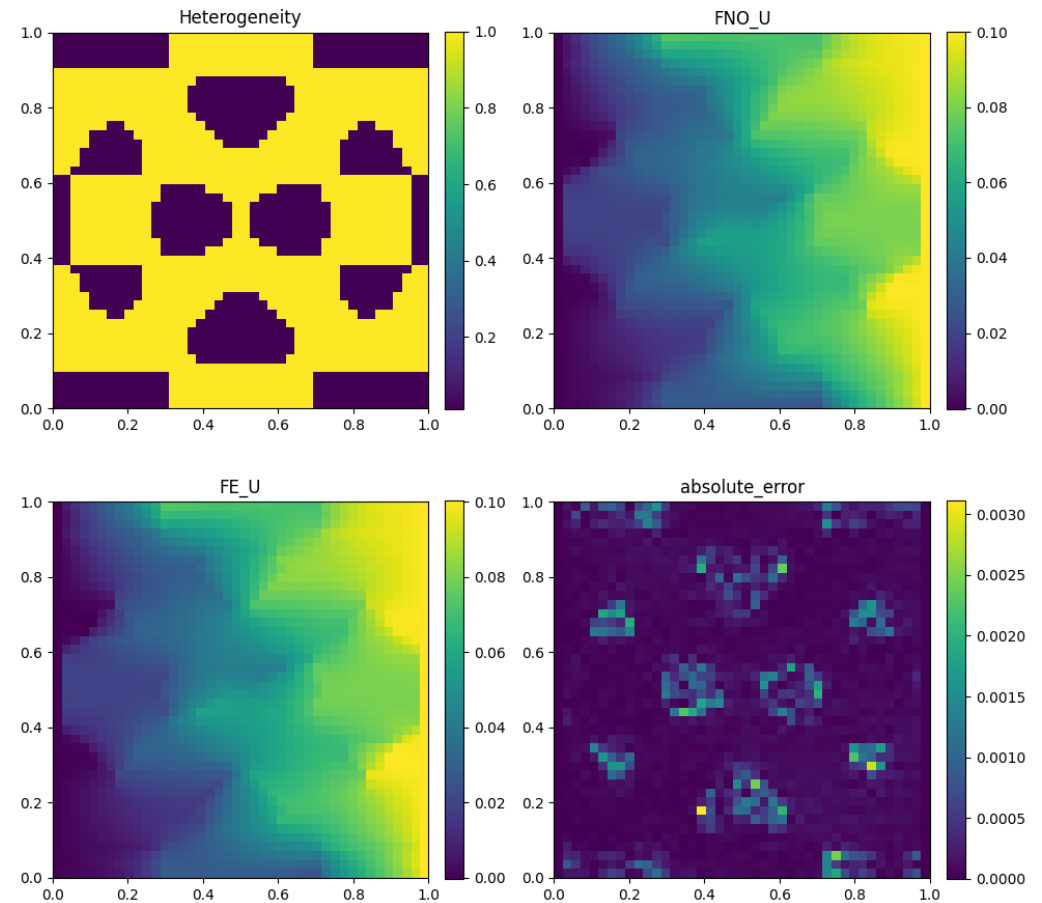
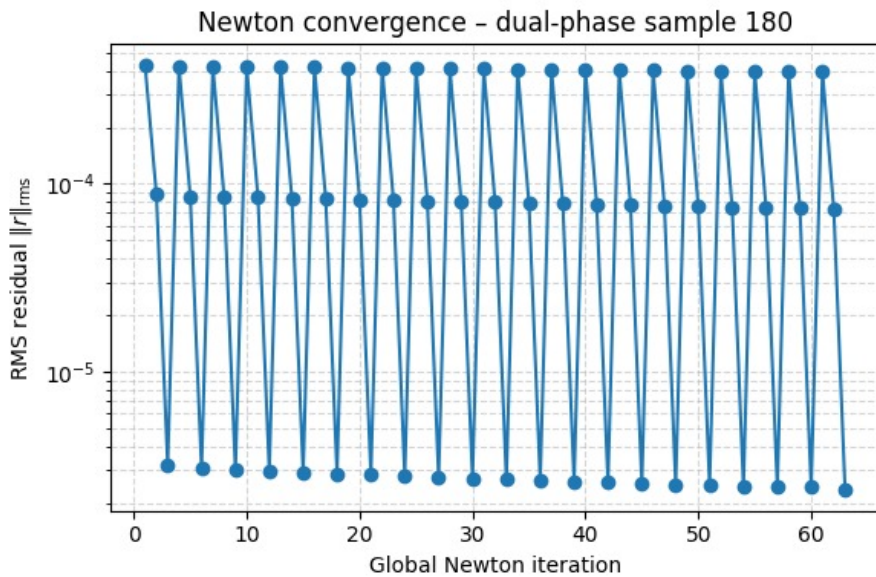


PARAMETRIC - TRANSFER LEARN - SLIDE 3.2

-Dual phase heterogeneous sample



7 mins for training — Auto-Stopped between 700-800 epochs



PARAMETRIC TRANSFER LEARN- SLIDE 3.3

-Dual phase heterogeneous sample with Circular inclusion

4 mins for training —Auto-Stopped between 500-600 epochs

