



DeepPore Dynamics

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Outline

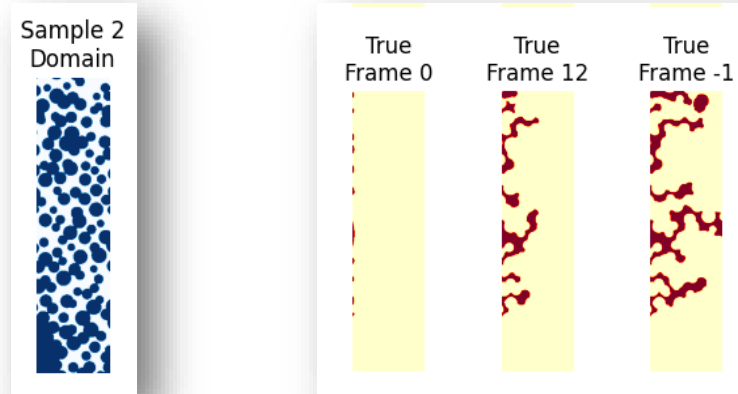
- Challenge and background
- Our approach – diffusion models
- Model parameters and results
- Iterative single time step iterations
- Next steps

The Challenge

- Building innovative AI models to predict CO2 invasion in a porous rock matrix.

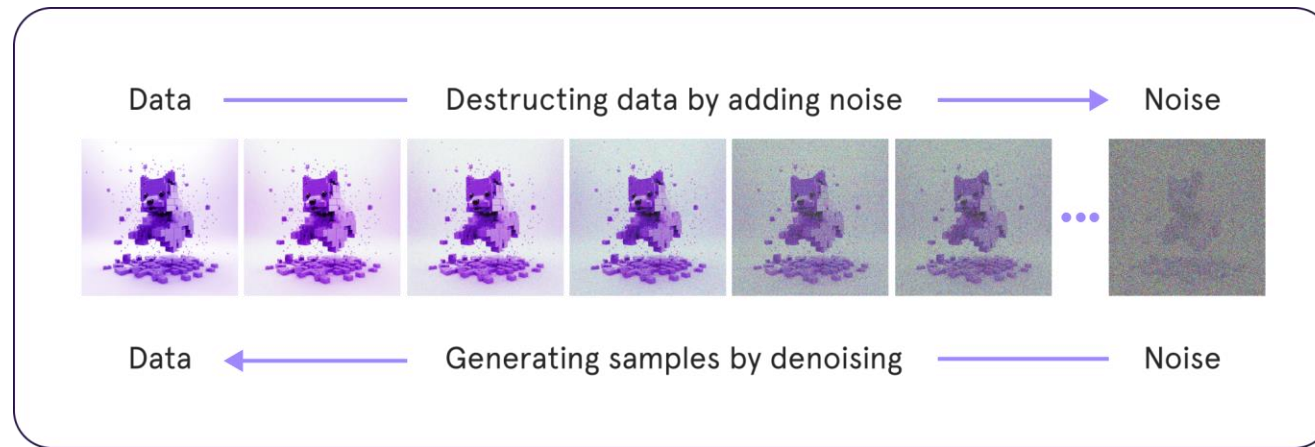
- Data:

- Domain arrays.
- Invasion arrays.



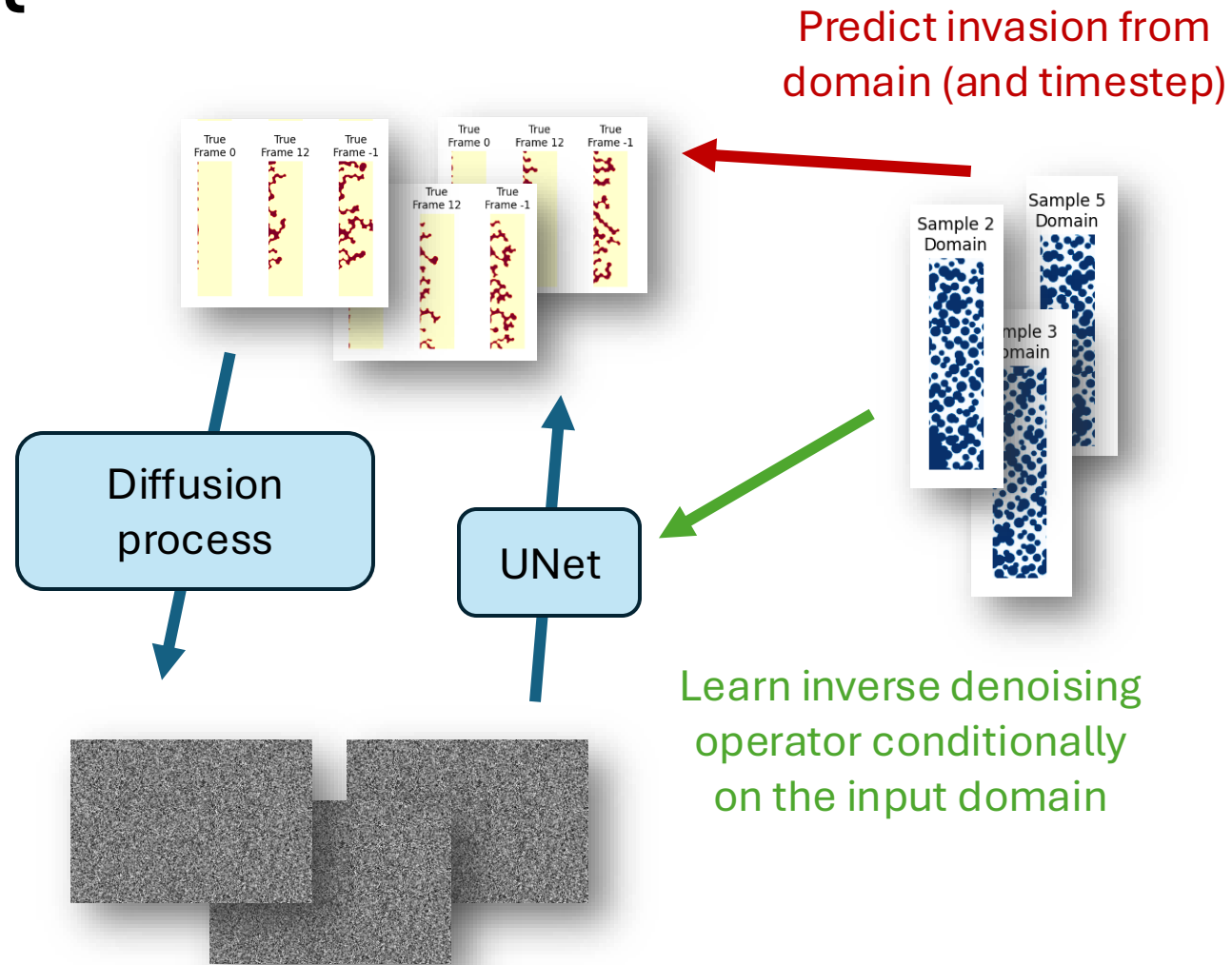
Diffusion models

- First introduced for image generation in Ho et al. (2020)
- Generative approach, outperforming GANs – Dhariwal et al. (2021)
- Work well for creating images from noise and a prompt



- Recent works have shown good results using diffusion models to predict fluid flow past an obstacle of varying shape – Hu et al. (2024)

The Model

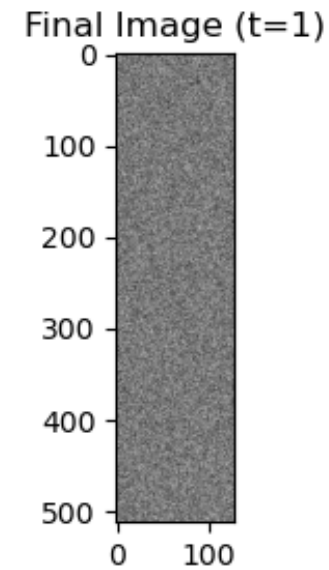
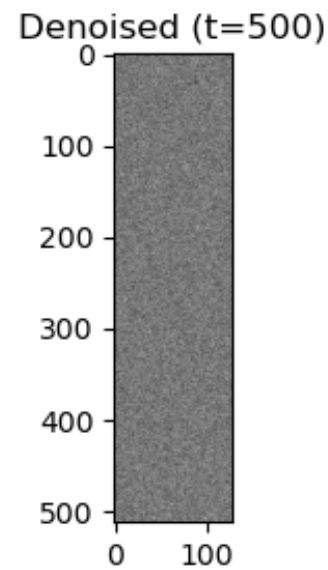
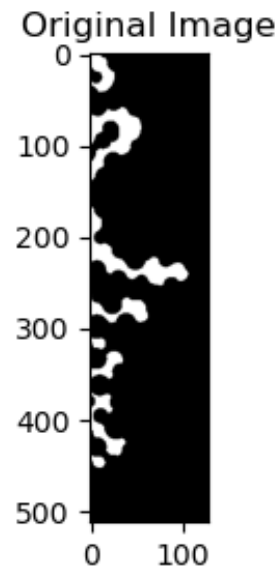


Diffusion Models Parameters

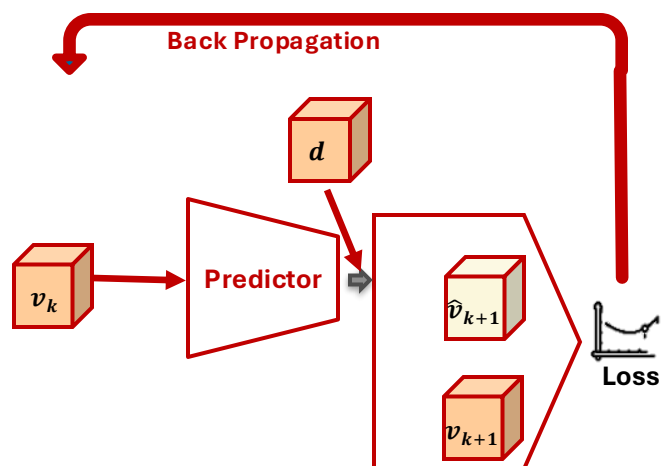
Time steps	Epochs	Data Size	Training time
1000	10	177	9 hrs

Results

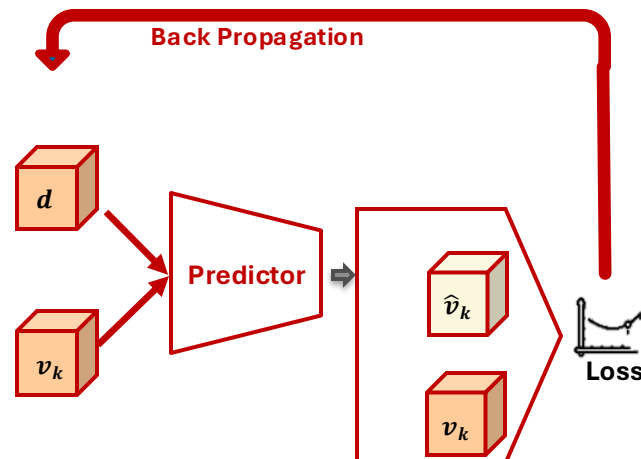
- The training loss ended with a MSE of 0.004
- The validation loss have an average value of about 0.0025.



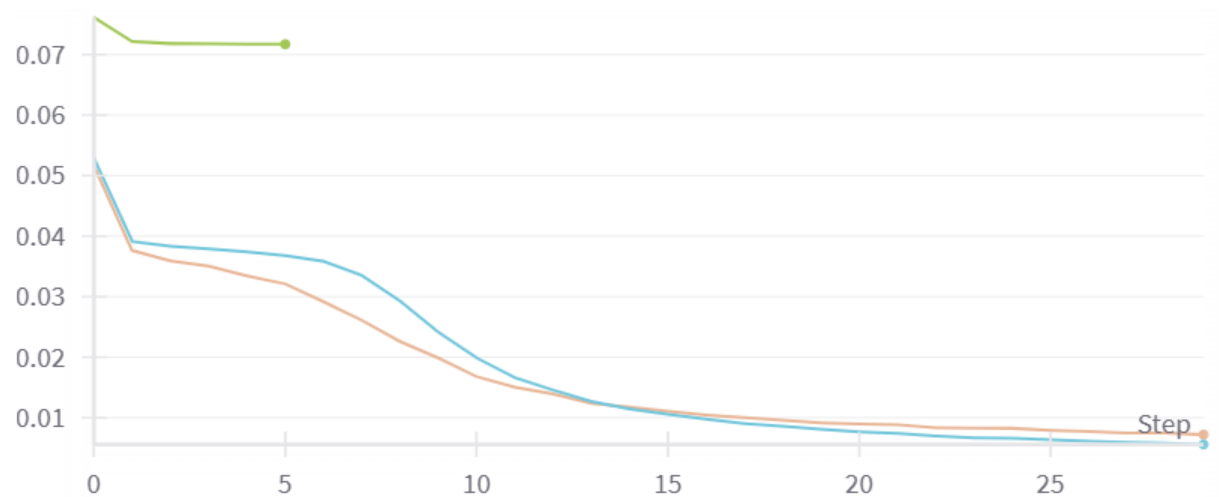
Unet explicit Prediction (training)



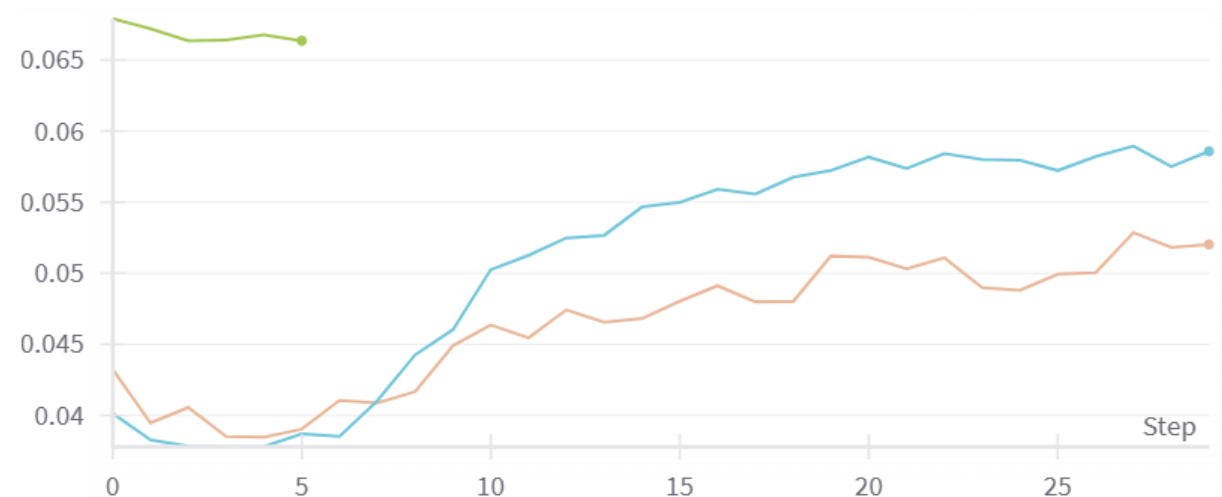
Unet implicit Prediction (training)



Training Loss

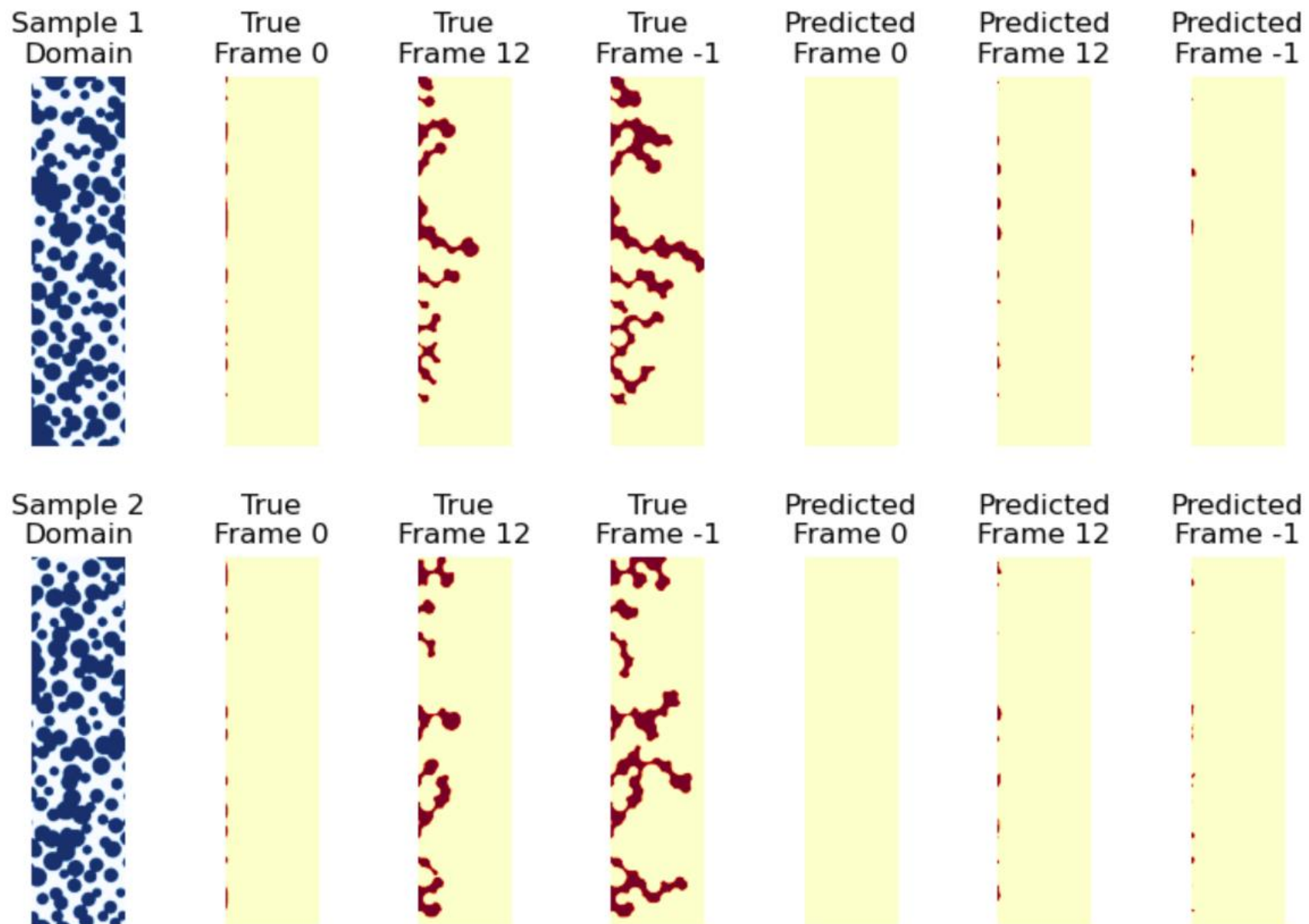


Testing Loss

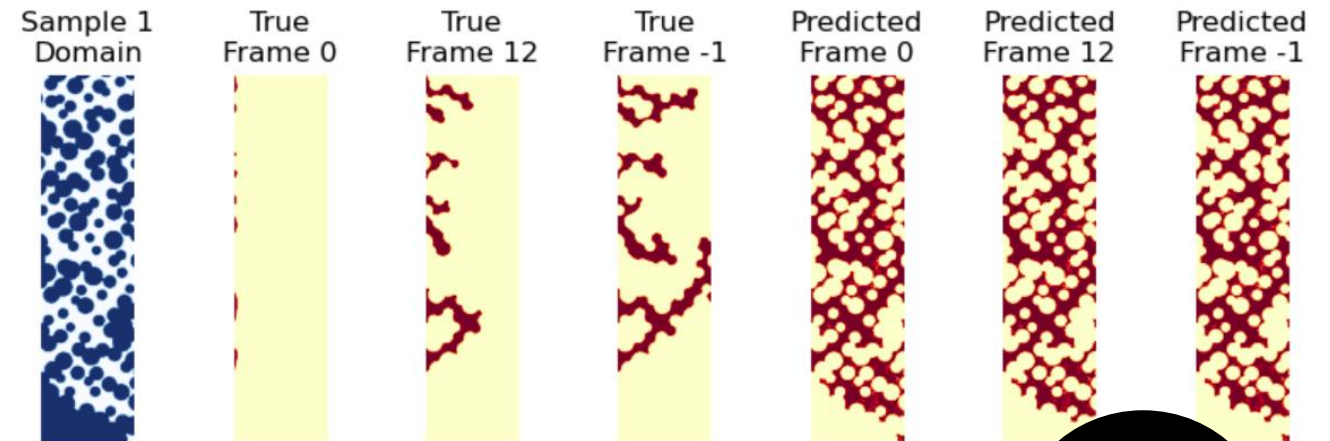
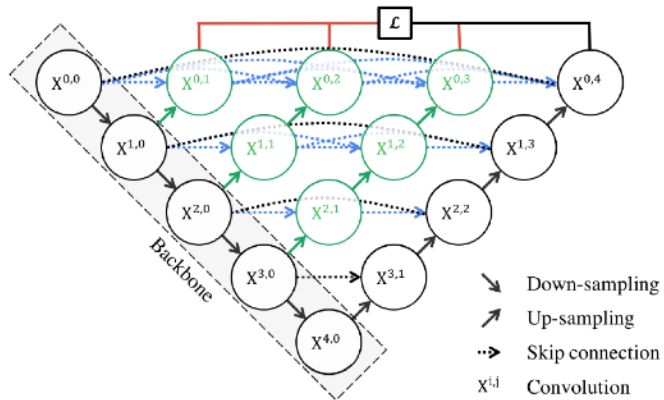
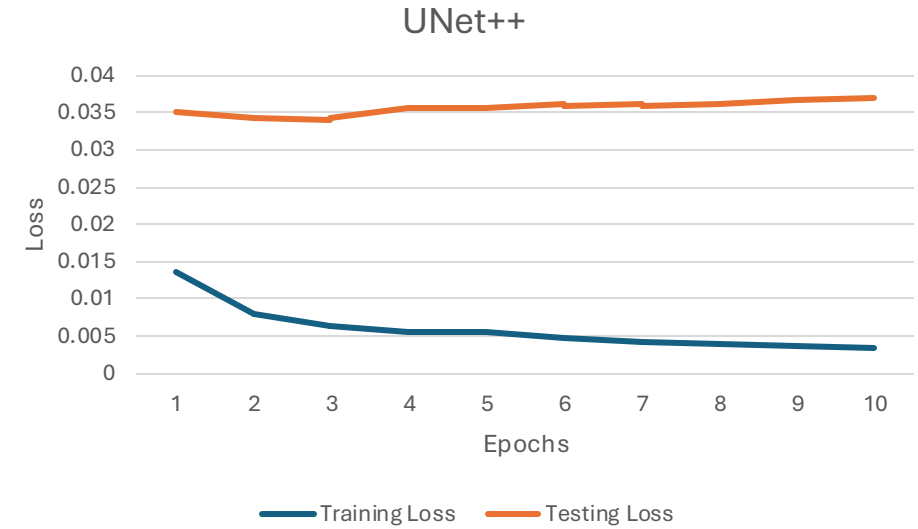
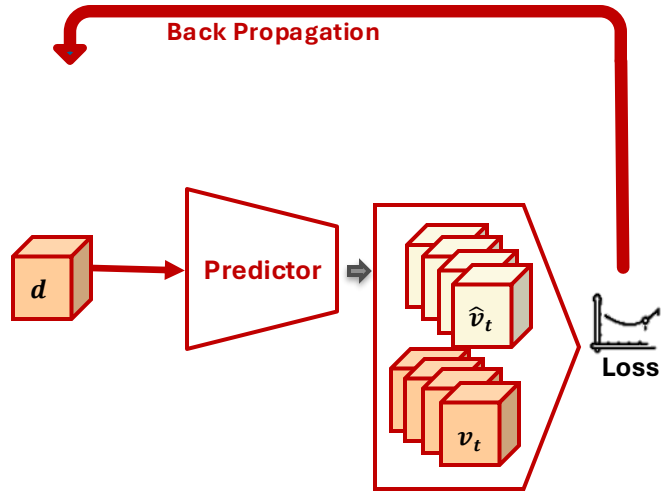


Unet explicit: 1 timestep input (previous t), domain as mask
 Unet explicit: 1 timestep input (previous t), domain as mask, subsamples
 Unet implicit: 2 inputs (guess t, domain)

Unet explicit: 1 timestep input (previous t), domain as mask



Unet++ Prediction – Multiple output (training)



Next steps

- Diffusion:
 - Correct the model and generate close denoised realizations.
 - Enhance the training by trying a different neural network.
 - Apply the domain data at the end to make sure that generated CO₂ are not located out of the pores.
- Assessing whether applying 3D Conv layers in the Unet generates better results, or adding LSTM instead of the Unet to integrate time sequence better.
- Try time embedding using Feature-wise Linear Modulation (FiLM)

$$y = \gamma(t) \cdot x + \beta(t)$$

Thank you

Appendix

