Neural Implicit Flow (NIF)

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Problem Statement

- Spatio-temporal data modeled by PDEs is computationally challenging.
- Current reduction methods (SVD, CAE) fail with variable geometry or adaptive meshing.
- Need for a scalable, mesh-agnostic approach for real-time engineering applications.
- Examples: Sea surface temperature modeling, turbulence modeling, sparse sensing.

New Approach: Neural Implicit Flow (NIF)

- Combines two neural networks:
 - ShapeNet: Encodes spatial complexity mesh-agnostically.
 - ParameterNet: Models temporal and parametric dependencies.
- Provides efficient, nonlinear dimensionality reduction and interpretable representations.

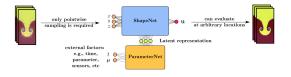


Figure: NIF Hypernetwork Architecture

My Contributions

- Performance tuning of the provided implementation.
- Re-Evaluation of the results
 - 40% better generalization performance
 - Scalable to complex spatio-temporal datasets

My Contributions - Performance tuning

- Performance tuning of the provided implementation.
 - Port from Tensorflow to PyTorch
 - Implementation using upstream SIREN
 - Improving parallelism in learning phase

My Contributions - Result Re-Evaluation

- Re-Evaluation of results with focus on chaotic systems
 - Comparison to other Hypernetworks / LoRA / etc
 - Comparing the existing results using own implementation