Solving Inverse Problems via Joint Modeling with Forward Process Improves Fidelity

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LLNL-PRES-XXXXX This work was performed under the auspices of the U.S. Department of Energy by Lawrence Livermore National Laboratory under contract DE-AC52-07NA27344. Lawrence Livermore National Security, LLC Transformer Architectures Provide a Unique Opportunity to Solve Inverse Problems

Conductivity field *x*

Inverse Problems:

Given only partially observed outputs can we recover the inputs that caused them?

Solution: Determine parameter distribution q(x|y) instead of inverse function (which does not exist).

Current deep learning based solutions do not model the forward and inverse processes jointly thus posing fidelity risks. Transformer architectures which act as unified architectures to map between sequences.

Forward Process



Data

Parameter 1

Inverse process



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Tasks

Regression Transformers For Posterior Inference



- Our transformer* architecture takes conductivity field and partially observed temperature field jointly as inputs:
- a) Interpolated measurement (based on 10 measurements); and b) A generated sample from the prior p(x).



experience, such as previous experiments.

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Generated solution

- Output of the transformer:
- Generated solutions to the inverse problem,
- Fully reconstructed measurement field.
- *Joint modeling* of prior sample and interpolated measurement to generate a solution.
- We designed novel loss functions to train this unified architecture.
- Once trained, our framework can generate all possible inverse solutions (conductivity fields) for any partially observed temperature field y.



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Born Jannis, Matteo Mancia. "Regression transformer enables concurrent sequence regression and generation for molecular language model", Nature Machine Intelligence 5.4 (2023):432-444.

Water Conductivity Field Inversion in a 2D Heat Equation



Solution to the inverse problem: Generated conductivity fields for a particular test measurement input







Let's check if our generated solutions are consistent!



Solution 1

Solution 2

Solution to the forward problem: Full temperature field reconstruction from generated conductivity fields

Solution 3







Black dots measured

MSE: 0.0048

GPR/Kriging



Proposed approach



Computed ECP Ideal ECP 0.8 ECP(q,α) ••• 0.2 0.0 0.0 0.4 10 Credibility level $(1-\alpha)$

Expected coverage probability (Lemos et al.) Our method consistently produces accurate solutions

We see that all the solutions produces back the same partially observed temperature field!

> Showing that we can recover all solutions (conductivity field) by our approach!

