







Geometric and Physical Constraints Synergistically Enhance Neural PDE Surrogates

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Project website



Scientific question?

Approaches to solving PDEs:

Numerical methods (traditional);

Machine Learning (ML) Models (recent)

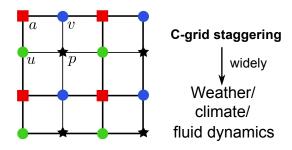
Challenges of ML model: Accurate long-term rollout; Generalization

PDEs fundamental assumptions: Symmetry and physical constraints

Previous ML Models

Related Works	symmetry constraint	physical constraint
Wandel et al. ICLR, 2021	×	V
Wang et al. ICLR, 2021	/	×

Incompressible
Navier–Stokes equation (INS)

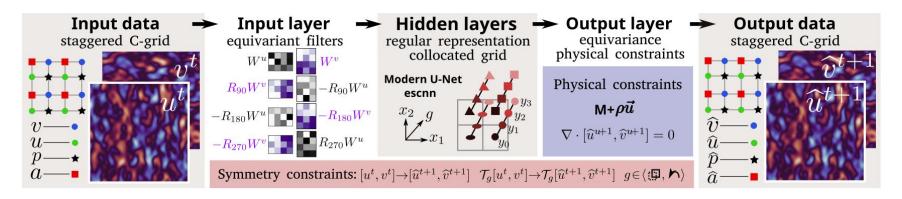


Our scientific question:

Would it be useful to combine the two constraints with the C-grid for the ML model, or would that be redundant?

Symmetry- and physics-constrained neural surrogate

Symmetry- and physics-constrained neural surrogate



Experiments

Shallow Water Equations (SWEs)

		Symmetries			
Conservation laws					
None Ø	p1/Ø	p4/Ø	p4m/∅		
Mass M	p1/M	p4/M	p4m/M		

Incompressible Navier–Stokes (INS) Decaying Turbulence

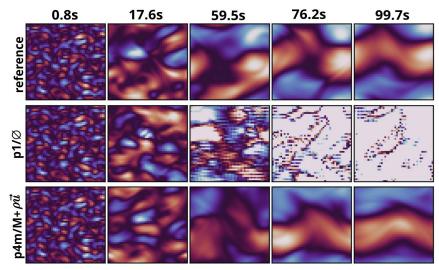
	Symmetries		
Conservation laws			回り単
None Ø	p1/Ø	p4/Ø	p4m/Ø
Momentum $\rho \vec{u}$	p1/ $ ho ec{u}$	p4/ $ ho ec{u}$	p4m/ $ ho ec{u}$
Mass/momentum M+ $\rho \vec{u}$	p1/M+ $ ho ec{u}$	p4/M+ $ ho ec{u}$	p4m/M+ $ ho ec{u}$

Our model outperforms other networks on SWEs and INS

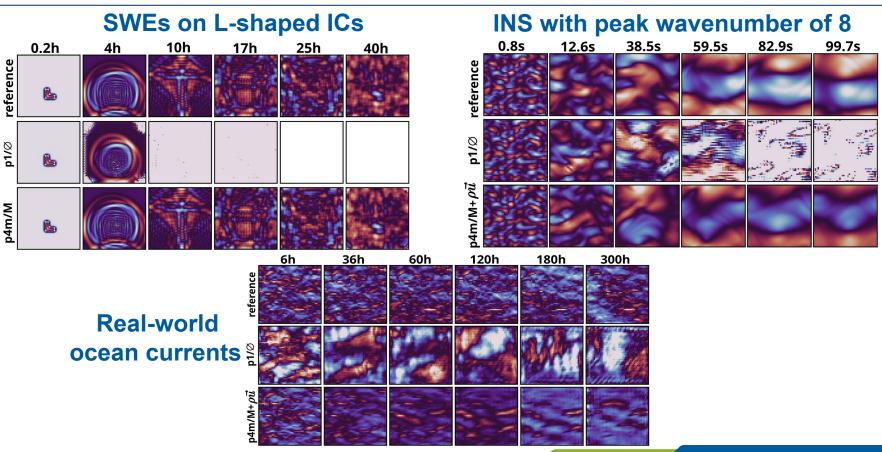
Prediction of p4m/M for SWEs

0.2h 2h 10h 25h 50h reference

Prediction of p4m/M+pu for INS



Generalization beyond training data



Conclusion

We developed a double-constrained model with input and output layers on C-grids.

We found that symmetries are more effective than physical constraints, but combining both is best.

We found our model can improve predictions in terms of generalisation and the real ocean data.

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