# **Experience Report of Physics-Informed Neural Networks in** Fluid Simulations: Pitfalls and Frustration

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### How Do PINNs (Physics-Informed Neural Networks) Work?

### Automatic differentiation



### Fig 1. Graphic illustration of PINNs

### **Study Overview**

### **Study Aim**

To understand the

- . feasibility of PINNs in practical engineering: controllability and predictability w.r.t. cost and accuracy, and
- 2. possibility of replacing traditional CFD solvers w/ PINNs.

### **Key Findings**

- 1. No obvious ways to control accuracy: no obvious translation from training loss to prediction errors
- 2. No obvious ways to predict time-to-solutions
- 3. Weak-scaling efficiency is good, but weak scaling does not help the accuracy nor the time-to-solution
- 4. Cost-performance ratio not competitive w/ traditional CFD solvers
- 5. Not able to solve a simple vortex-shedding problem

### **Limitations and Disclaimer**

- 1. We only consider the data-free applications of PINNs.
- 2. We did not exhaust all possible architectures and configurations. The qualitative findings only apply to the specific configurations we tried.

### Acknowledgement

We appreciate the support by NVIDIA, through sponsoring the access to its high-performance computing cluster.

### Multiple objectives

## Analytica 2.5 -2.5-0.48 - 0.24 0.00 0.24 0.48Analytical -2.52.5 0.0

-0.36 -0.24 -0.12 0.00 0.12 0.24



### Vortex Shedding Benchmarking: PINNs Gave Steady-State Solution



Fig 7. Finite difference solution: flow field is expected to have vortex shedding



### Cost-Performance Benchmarking w/ a 2D Taylor-Green Vortex Problem



### Fig 2. Contours showing degraded temporal accuracy

Fig 3. Comparing the cost-performance behaviors against traditional CFD solvers



### Fig 4. Loss and accuracy versus training iterations show no straightforward relationship



Fig 5. Accuracy versus wall time shows no clear relationship between batch size and time-to-solution

Fig 8. PINNs simulation: no shedding at all

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Fig 6 and Table 1. Weak-scaling benchmarking. No obvious benefit to convergence.

Fig 9. Drag and lift coefficients. PINNs results show a steady-state behavior.