

# Kartikey Singh

Undergraduate Researcher /  
Scientific Machine Learning &  
PDEs

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Portfolio | Published Paper

## Summary

Undergraduate researcher who **discovered and formalized Solution Manifold Incompatibility**—a fundamental structural flaw in linear multi-fidelity physics-informed neural networks. The work followed an **implementation-first research approach**: building and stress-testing multi-fidelity PINN systems, isolating systematic failure under linear fusion, and subsequently formalizing the observed behavior via operator non-closure and tangency arguments. Validated findings using PyTorch PINNs and a custom high-performance CFD solver benchmarked to  $Re=1000$ . Published independent paper with permanent DOI and open-sourced 5000+ lines of reproducible research code. Demonstrated ability to identify core failure modes, design diagnostics, and translate computational observations into rigorous theory.

**Context:** Conducted independently prior to the formal research year of the undergraduate program.

**Research approach:** Implementation-driven discovery — building computational experiments to reveal fundamental principles, then formalizing insights mathematically.

## Research Experience

Sep 2025–Present	Independent Research: Solution Manifold Incompatibility
	<ul style="list-style-type: none"> <li>○ <b>Implementation-Driven Exploration</b> (Sep–Oct 2025):           <ul style="list-style-type: none"> <li>- Built and trained baseline multi-fidelity PINN implementations for nonlinear PDEs</li> <li>- Observed systematic degradation under linear solution fusion despite accurate, independently validated single-fidelity models</li> <li>- Isolated failure to the fusion mechanism via freeze-and-blend experiments</li> </ul> </li> <li>○ <b>Formalization, Diagnostics &amp; Validation</b> (Oct–Nov 2025):           <ul style="list-style-type: none"> <li>- Formalized Solution Manifold Incompatibility as operator non-closure under linear blending</li> <li>- Derived nonlinear cross-interaction expansions and tangency failure results</li> <li>- Designed structural diagnostics (<math>C_{NL}</math>, blending activity index)</li> <li>- Validated findings via PyTorch PINNs and custom CFD solver (Re=1000)</li> </ul> </li> <li>○ <b>Dissemination</b> (Late 2025–Present):           <ul style="list-style-type: none"> <li>- Published paper with permanent DOI (Zenodo)</li> <li>- Open-sourced 5000+ LOC codebase</li> <li>- Defended research before academic panel</li> </ul> </li> </ul>
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	<b>Publications</b>
2025	<p><b>Solution Manifold Incompatibility in Multi-Fidelity PINNs, K. Singh</b></p> <ul style="list-style-type: none"> <li>○ Zenodo (preprint), 35 pages</li> <li>○ Introduces the SMI framework with operator-level analytical proofs</li> <li>○ Presents numerical experiments across three PDE classes</li> <li>○ Open-source code: <a href="https://github.com/KartikeyGangwar/MF-PINN-SMI-Failure-Modes">https://github.com/KartikeyGangwar/MF-PINN-SMI-Failure-Modes</a></li> </ul>
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	<b>Technical Skills</b>
Machine Learning & AI	Physics-Informed Neural Networks, Multi-fidelity learning, PyTorch (automatic differentiation, custom training loops), Neural operator architectures (theoretical exposure), Research code development
Scientific Computing & High Performance Computation	High-performance CFD (Finite Difference Methods: ADI, SOR), Streamfunction–Vorticity formulation, Vectorized NumPy, Convergence analysis, Benchmark validation (to Re=1000), GPU-accelerated PINN training with PyTorch (CUDA)

Mathematical Foundations	Nonlinear PDE theory, Numerical analysis, Solution manifold geometry, Functional analysis, Differential geometry, Operator theory
Software Development & Tools	Python (PyTorch, NumPy, SciPy, Matplotlib), LaTeX, Git/GitHub, Linux, VS Code, Research pipeline automation, Documentation
Research Communication	Publication-quality visualization, Technical writing, Research presentations, Academic defense, Open-source documentation, Reproducible workflows
Research Methodology	Implementation-first research, Empirical analysis of numerical methods, Debugging complex computational systems, Translating observations to theory

## Software & Research Tools

- 2025–2026 **Open-Source Research Suite: SMI Study**
- **Purpose:** Demonstrate Solution Manifold Incompatibility in multi-fidelity PINNs
  - **Implementation:** Complete PyTorch pipeline for three PDE families (Allen–Cahn, Burgers, Navier–Stokes)
  - **Features:** Automated diagnostics ( $C_{NL}$ ), multi-level documentation, reproducibility protocols
  - **Repository:** GitHub | **DOI:** 10.5281/zenodo.17794638
- 2025–2026 **Physics-Informed Neural Networks for Fluid Dynamics**
- Implemented PINN solvers for Burgers' equation and 2D incompressible Navier–Stokes
  - Studied formulation sensitivity at moderate-to-high Reynolds numbers ( $Re \approx 10^3$ )
  - Compared streamfunction–vorticity ( $\psi-\omega$ ) and streamfunction–pressure ( $\psi-p$ ) formulations
  - Demonstrated failure modes of data-free and physics-dominated PINNs in cavity flow
  - Showed superior robustness of  $\psi-p$  formulation without explicit wall-vorticity closure
  - Emphasized transparent documentation of failure modes over aggressive tuning
  - **Repository:** GitHub
- 2025 **High-Performance CFD Solver**
- **Purpose:** Generate high-fidelity ground truth for ML validation
  - **Methods:** Streamfunction–Vorticity formulation with ADI scheme (vorticity) and Red-Black SOR (Poisson)
  - **Performance:** Validated solutions up to **Re=1000** on  $251 \times 251$  grids, matching classical benchmarks
  - **Repository:** GitHub | **DOI:** 10.5281/zenodo.18312938

## Education

2023–2027 **B.Sc. (Honors) Mathematics**, *University of Delhi*, Delhi, India  
(Expected)

- **Program Structure:** 4-year undergraduate program with a formal research year (2026–2027)
- **Relevant Coursework:** Real Analysis, Differential Equations, Numerical Methods, Complex Analysis, Linear Algebra
- **Independent Research:**
  - Self-directed study in Physics-Informed Machine Learning (2024–2025)
  - Advanced PDE theory and scientific computing (2024–present)
  - Developed complete research pipeline outside formal curriculum
- **Achievements:**
  - Published independent research paper as undergraduate
  - Developed and open-sourced research software with DOIs
  - Successfully defended research before professor panel

## Additional Information

Languages	English (Professional proficiency), Hindi (Native)
Technical Interests	Scientific machine learning, Nonlinear PDEs, High-performance computing, Open-source scientific software
Research Philosophy	Belief in reproducible research, open science, and bridging theory with practical implementation