

## AUTHORS

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# PINA: a PyTorch Framework for Solving Differential Equations by Deep Learning for Research and Production Environments

PINA is an open-source software powered by PyTorch and Lightning, designed for solving differential equations with neural networks. It supports Physics Informed Neural Networks and Neural Operators, offering flexibility for users to craft models tailored to their needs. PINA is modular and adaptable to different hardware setups, including GPUs and TPUs.

## AFFILIATIONS

SISSA mathLab, Trieste, Italy  
FAST Computing Srl, Trieste, Italy

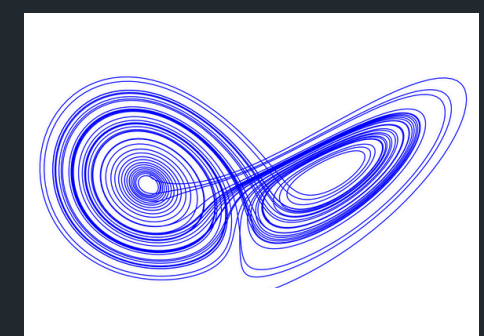


## SOLVE YOUR DIFFERENTIAL EQUATION STEP BY STEP

$$\begin{cases} \dot{x} = \sigma(y - x) \\ \dot{y} = x(\rho - z) - y \\ \dot{z} = xu - \beta z \end{cases}$$

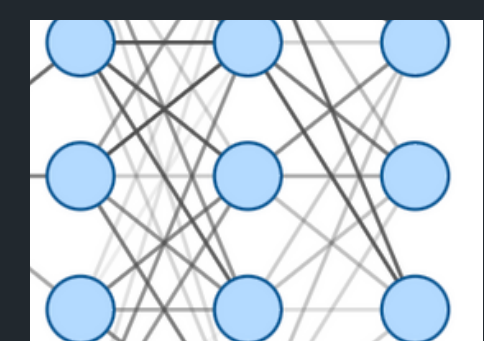
### Problem definition

Specify the mathematical problem aimed to be solved and the specific physical condition to be satisfied



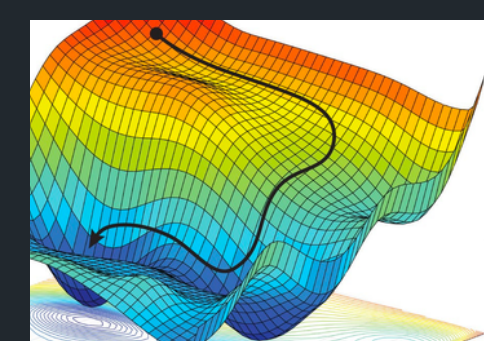
### Sample the domain

Prepare the input of the model by discretising the physical domain, or import data from numerical solvers



### Model & Solver selection

Build a Model as a PyTorch Module and choose the Solver strategy to optimize the model and solve the problem



### Training

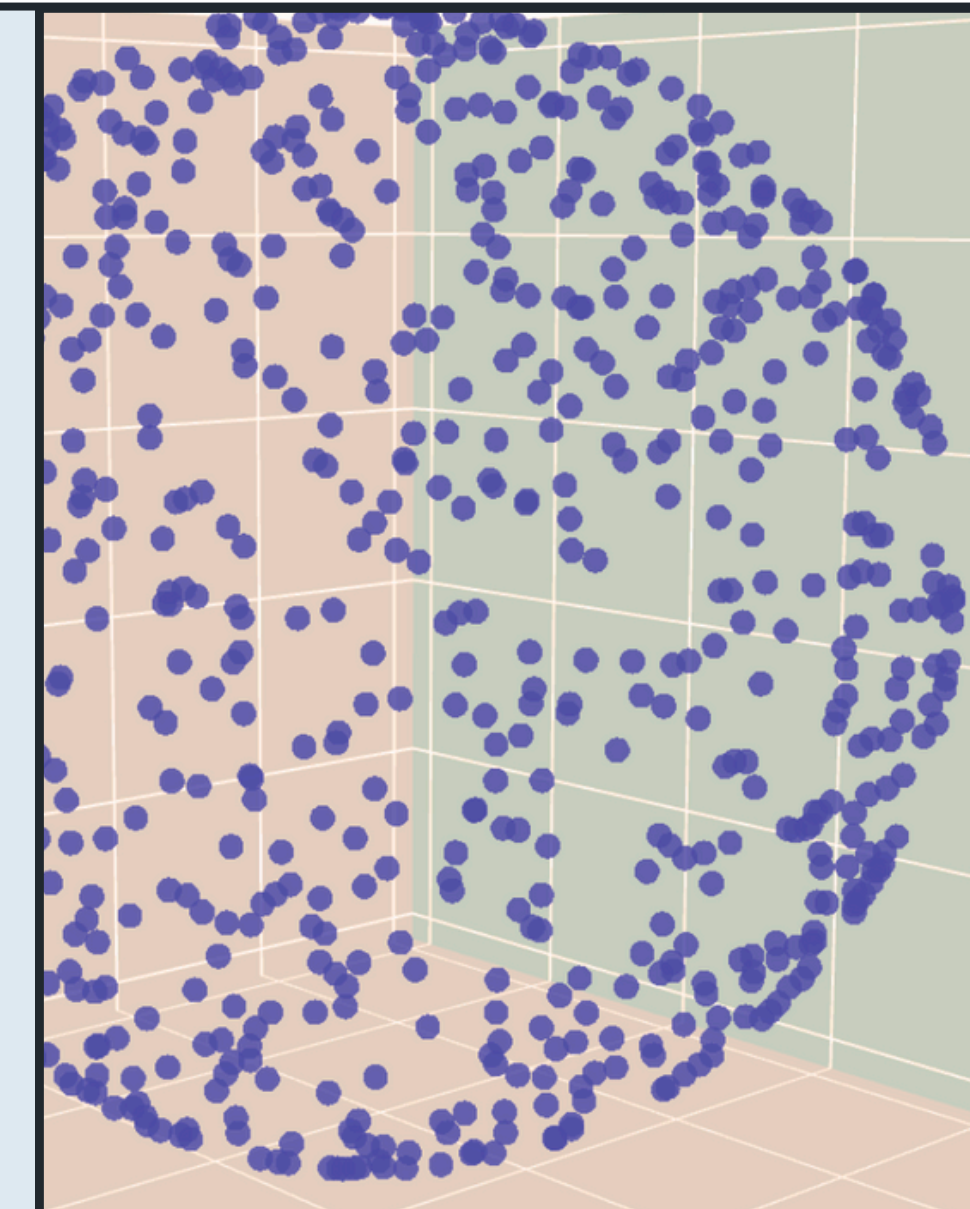
Optimize the Model with the specific Solver strategy with all the additional Pytorch Lightning features

## TRASLATE YOUR PROBLEM INTO PINA LANGUAGE

```
class Poisson(SpatialProblem):
    output_variables = ['u']
    spatial_domain = CartesianDomain({'x': [0, 1], 'y': [0, 1]})

    conditions = {
        'gamma1': Condition(
            location=CartesianDomain({'x': [0, 1], 'y': [1]}),
            equation=FixedValue(0.0)),
        ...
        'gamma4': Condition(
            location=CartesianDomain({'x': 0, 'y': [0, 1]}),
            equation=FixedValue(0.0)),
        'D': Condition(
            location=CartesianDomain({'x': [0, 1], 'y': [0, 1]}),
            equation=my_laplace),
        'data': Condition(input_points=in_, output_points=out_)
    }
```

- Spatial, Time-dependent, Inverse, and Parametric problems are already available!
- We implemented the most common differential operators for easiness of usage.
  - grad
  - div
  - laplacian
- Triangular, cartesian, and elliptic shapes can be used to define the problem domain. Moreover, boolean operations have been implemented to make possible the definition of complex domains!



## A HIERACHICAL PERSPECTIVE: SOLVER, MODEL, LAYER

PINA makes available the latest methodologies for equation learning. To maximize flexibility and modularity, the methodological implementation is divided into Solver, Model, and Layer

### Solver

The “strategy” we want to apply to solve the problem  
--- e.g. PINN

- Physics-informed paradigm, the solver minimizes the physical residual
  - PINN, GPINN, CausalPINN, CompetitivePINN, SAPINN, ...
- Supervised learning paradigm, the solver minimizes the difference between data and network output
  - SupervisedSolver, GAROM, MessagePassingNeuralPDE, ReducedOrderModellingSolver ...
- Abstract Interfaces to easily build new solvers
  - SolverInterface, PINNInterface, ...

### Model

The architecture to apply  
--- e.g. DeepONet

- Standard and customizable deep learning architectures
  - FeedForward, MultiFeedForward, ResidualFeedForward, ...
- Specific Neural Operator architectures
  - FourierNeuralOperator, LowRankNeuralOperator, AveragingNeuralOperator, MIONet, DeepONet, ...
- Easily build your PytorchModel or use our abstract Interfaces
  - KernelNeuralOperator, Network, ...

### Layer

The minimal brick that can be used to build a Model  
--- e.g. SpectralConv

- Many PyTorch implementations of deep learning Layers
  - ContinuousConvBlock, ResidualBlock, EnhancedLinear
  - SpectralConvBlock1D, SpectralConvBlock2D, SpectralConvBlock3D,
  - FourierBlock1D, FourierBlock2D, FourierBlock3D,
  - POBlock, PeriodicBoundaryEmbedding, AVNOBlock, LowRankBlock
- Adaptive Activation Functions

## HIGHLIGHTS

### PINA is built upon PyTorch Lightning

- CPU GPU and TPU training support
- Loggers and Checkpoints for monitoring training
- Gradient Clipping, SWA, Gradient accumulation ...
- Callbacks for Solvers and Trainer

### LabelTensor

- Extension of PyTorch Tensor class to handle labels
- Easily extract variables with strings and compute differential operator in symbolic notation
- Compatible with PyTorch main Tensor operations



x	y	z	t
2.3	4.5	1.2	0.2
3.7	6.1	9.8	1.3
5.2	7.4	3.9	4.5
8.6	2.4	6.7	2.9
1.5	9.3	4.8	6.1

### High-level design

- Modular components and object-oriented structure
- Abstract interfaces to easily add new components
- Natively support PyTorch Model compilation



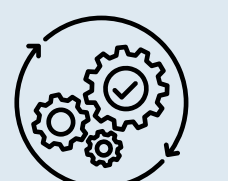
### Documentation online

- Fully documented package
- Installation and contribution guidelines
- Tutorials for getting start with the software



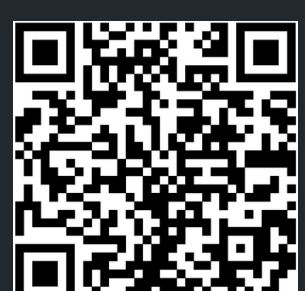
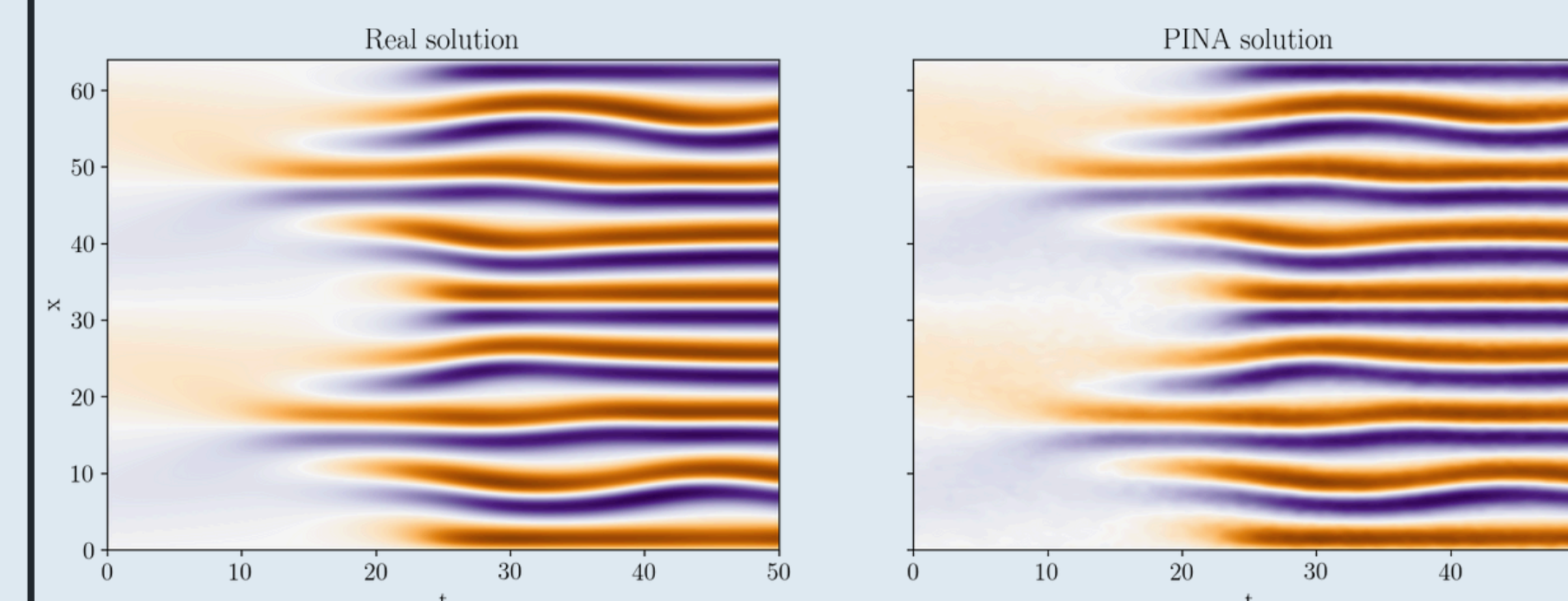
### Tested for several operating systems

- Fully compatible and tested for Python ≥ 3.7
- Running on Windows macOS and Ubuntu



### Easy installation

- pip install pina-mathlab
- New release the 1st of any month



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